

## Discrete Wavelet Transforms Based on Zero-Phase Daubechies Filters

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overheads for talk available at

<http://faculty.washington.edu/dbp/talks.html>

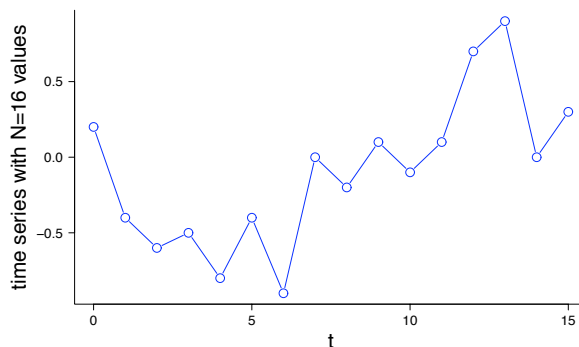
## Overview

- will discuss work in progress on the ‘zephlet’ transform, an orthonormal discrete wavelet transform (DWT) based on zero-phase filters
- will start by giving some background on the DWT as formulated in Daubechies (1992) – see, e.g., Percival & Walden (2000) or Gençay et al. (2002) for further details
- will then describe the zephlet transform and how it differs from the usual DWT, with an illustration of some of its properties

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## Background on DWT: I

- let  $\mathbf{X} = [X_0, X_1, \dots, X_{N-1}]^T$  be a vector of  $N$  time series values (note: ‘ $T$ ’ denotes transpose; i.e.,  $\mathbf{X}$  is a column vector)
- for simplicity, assume  $N$  is an even number



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## Background on DWT: II

- DWT is a linear transform of  $\mathbf{X}$  yielding  $N$  DWT coefficients
- notation:  $\mathbf{W} = \mathcal{W}\mathbf{X}$ , where  $\mathbf{W}$  is vector of DWT coefficients, and  $\mathcal{W}$  is  $N \times N$  *orthonormal* transform matrix
- orthonormality says  $\mathcal{W}^T\mathcal{W} = I_N$  ( $N \times N$  identity matrix)
- orthonormality is exploited heavily in, among other uses, DWT-based extraction of signals (‘wavelet shrinkage’)
- to focus discussion, will concentrate on so-called unit-level DWT, for which  $\mathbf{W} = [\mathbf{W}_1^T, \mathbf{V}_1^T]^T$ , where the two subvectors contain
  - wavelet coefficients  $\mathbf{W}_1 = [W_{1,0}, W_{1,0}, \dots, W_{1, \frac{N}{2}-1}]^T$  and
  - scaling coefficients  $\mathbf{V}_1 = [V_{1,0}, V_{1,0}, \dots, V_{1, \frac{N}{2}-1}]^T$
- higher-level DWTs use unit-level DWTs over and over again

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### The Wavelet Filter: I

- matrix  $\mathcal{W}$  is rarely constructed explicitly, but rather is formed implicitly by use of a wavelet filter
- let  $\{h_l : l = 0, \dots, L - 1\}$  be a real-valued filter of width  $L$
- for convenience, will define  $h_l = 0$  for  $l < 0$  and  $l \geq L$

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### The Wavelet Filter: II

- $\{h_l\}$  called a wavelet filter if it has these 3 properties

1. summation to zero:

$$\sum_{l=0}^{L-1} h_l = 0$$

2. unit 'energy' (i.e., squared Euclidean norm):

$$\sum_{l=0}^{L-1} h_l^2 = 1$$

3. orthogonality to even shifts: for all nonzero integers  $n$ , have

$$\sum_{l=0}^{L-1} h_l h_{l+2n} = 0$$

- 2 and 3 together are called the *orthonormality property*

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### The Wavelet Filter: III

- summation to zero and unit energy relatively easy to achieve
- orthogonality to even shifts is key property & hardest to satisfy (implies  $L$  must be even; common choices are 2, 4, ..., 20)
- define transfer function for wavelet filter, i.e., its discrete Fourier transform (DFT), along with its squared gain function:

$$H(f) \equiv \sum_{l=0}^{L-1} h_l e^{-i2\pi fl} \quad \text{and} \quad \mathcal{H}(f) \equiv |H(f)|^2$$

- orthonormality property is equivalent to

$$\mathcal{H}(f) + \mathcal{H}(f + \frac{1}{2}) = 2 \quad \text{for all } f$$

(an elegant – but not obvious! – result)

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### The Wavelet Filter: IV

- simplest wavelet filter is Haar ( $L = 2$ ):  $h_0 = \frac{1}{\sqrt{2}}$  &  $h_1 = -\frac{1}{\sqrt{2}}$
- note that  $h_0 + h_1 = 0$  and  $h_0^2 + h_1^2 = 1$ , as required
- orthogonality to even shifts also readily apparent
- squared gain function is

$$\mathcal{H}(f) = 2 \sin^2(\pi f),$$

for which

$$\begin{aligned} \mathcal{H}(f) + \mathcal{H}(f + \frac{1}{2}) &= 2 \sin^2(\pi f) + 2 \sin^2(\pi[f + \frac{1}{2}]) \\ &= 2 \sin^2(\pi f) + 2 \cos^2(\pi f) \\ &= 2, \end{aligned}$$

as required

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## Construction of Wavelet Coefficients: I

- given wavelet filter  $\{h_l\}$  of width  $L$  & time series of even length, obtain wavelet coefficients as follows

- circularly* filter  $\mathbf{X}$  with wavelet filter to yield output

$$\sum_{l=0}^{L-1} h_l X_{t-l} = \sum_{l=0}^{L-1} h_l X_{t-l \bmod N}, \quad t = 0, \dots, N-1;$$

i.e., if  $t-l$  does not satisfy  $0 \leq t-l \leq N-1$ , interpret  $X_{t-l}$  as  $X_{t-l \bmod N}$ ; for example,  $X_{-1} = X_{N-1}$  and  $X_{-2} = X_{N-2}$

- take every other value of filter output to define

$$W_{1,t} \equiv \sum_{l=0}^{L-1} h_l X_{2t+1-l \bmod N}, \quad t = 0, \dots, \frac{N}{2} - 1;$$

$\mathbf{W}_1$  formed by *downsampling* filter output by a factor of 2

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## Construction of Wavelet Coefficients: II

- can write  $\mathbf{W}_1 = \mathcal{W}_1 \mathbf{X}$ , where, when  $N \geq 10$  for example,

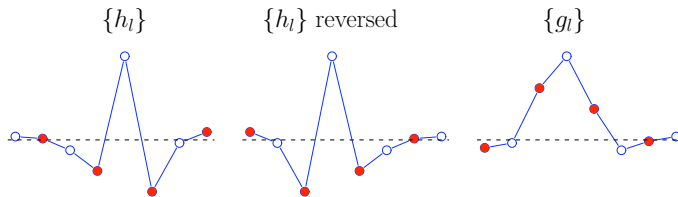
$$\mathcal{W}_1 \equiv \begin{bmatrix} h_1^\circ & h_0^\circ & h_{N-1}^\circ & h_{N-2}^\circ & h_{N-3}^\circ & \cdots & h_5^\circ & h_4^\circ & h_3^\circ & h_2^\circ \\ h_3^\circ & h_2^\circ & h_1^\circ & h_0^\circ & h_{N-1}^\circ & \cdots & h_7^\circ & h_6^\circ & h_5^\circ & h_4^\circ \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots \\ h_{N-1}^\circ & h_{N-2}^\circ & h_{N-3}^\circ & h_{N-4}^\circ & h_{N-5}^\circ & \cdots & h_3^\circ & h_2^\circ & h_1^\circ & h_0^\circ \end{bmatrix}$$

- here  $h_l^\circ = h_l$  when  $L \leq N$ , but takes different form if  $L > N$ ; for example, if  $N = 10$  and  $L = 20$ ,  $h_l^\circ = h_l + h_{l+10}$
- can argue that  $\mathcal{W}_1 \mathcal{W}_1^T = I_{N/2}$  for all  $L$  and  $N$
- $\mathcal{W}_1$  is the top *half* of orthonormal transform matrix  $\mathcal{W}$

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## The Scaling Filter: I

- create scaling filter  $\{g_l\}$  by reversing  $\{h_l\}$  and then changing sign of coefficients with even indices



- precise definition is  $g_l \equiv (-1)^{l+1} h_{L-1-l}$

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## The Scaling Filter: II

- properties 2 and 3 (orthonormality) of  $\{h_l\}$  are shared by  $\{g_l\}$ :
  - unit energy:

$$\sum_{l=0}^{L-1} g_l^2 = 1$$

- orthogonality to even shifts: for all nonzero integers  $n$ , have

$$\sum_{l=0}^{L-1} g_l g_{l+2n} = 0$$

- squared gain function  $\mathcal{G}(\cdot)$  for scaling filter satisfies

$$\mathcal{G}(f) = \mathcal{H}(f + \frac{1}{2}) \text{ and hence } \mathcal{H}(f) + \mathcal{G}(f) = 2$$

is equivalent way of stating orthonormality property

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## Construction of Scaling Coefficients: I

- orthonormality property of  $\{h_l\}$  is all that is needed to prove  $\mathcal{W}_1$  is half of an orthonormal transform (never used  $\sum_l h_l = 0$ )
- going back and replacing  $h_l$  with  $g_l$  everywhere yields another half of an orthonormal transform
- circularly filter  $\mathbf{X}$  using  $\{g_l\}$  and downsample to define scaling coefficients:

$$V_{1,t} \equiv \sum_{l=0}^{L-1} g_l X_{2t+1-l \bmod N}, \quad t = 0, \dots, \frac{N}{2} - 1$$

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## Construction of Scaling Coefficients: II

- have  $\mathbf{V}_1 = \mathcal{V}_1 \mathbf{X}$ , where  $\mathcal{V}_1$  is analogous to  $\mathcal{W}_1$ :

$$\mathcal{V}_1 = \begin{bmatrix} g_1^\circ & g_0^\circ & g_{N-1}^\circ & g_{N-2}^\circ & g_{N-3}^\circ & \cdots & g_5^\circ & g_4^\circ & g_3^\circ & g_2^\circ \\ g_3^\circ & g_2^\circ & g_1^\circ & g_0^\circ & g_{N-1}^\circ & \cdots & g_7^\circ & g_6^\circ & g_5^\circ & g_4^\circ \\ \vdots & \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots & \vdots \\ g_{N-1}^\circ & g_{N-2}^\circ & g_{N-3}^\circ & g_{N-4}^\circ & g_{N-5}^\circ & \cdots & g_3^\circ & g_2^\circ & g_1^\circ & g_0^\circ \end{bmatrix}$$

- as before, can argue that  $\mathcal{V}_1 \mathcal{V}_1^T = I_{N/2}$
- in addition, each row in  $\mathcal{W}_1$  is orthogonal to each row in  $\mathcal{V}_1$  and hence

$$\mathcal{W} \equiv \begin{bmatrix} \mathcal{W}_1 \\ \mathcal{V}_1 \end{bmatrix} \text{ is an orthonormal transform}$$

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## Daubechies Scaling Filters

- Daubechies (1992) constructs a family of scaling filters  $\{g_l\}$  with squared gain functions given by

$$\mathcal{G}_{(D)}(f) \equiv 2 \cos^L(\pi f) \sum_{l=0}^{\frac{L}{2}-1} \binom{\frac{L}{2}-1+l}{l} \sin^{2l}(\pi f)$$

(corresponding wavelet filter given by  $h_l = (-1)^l g_{L-1-l}$ )

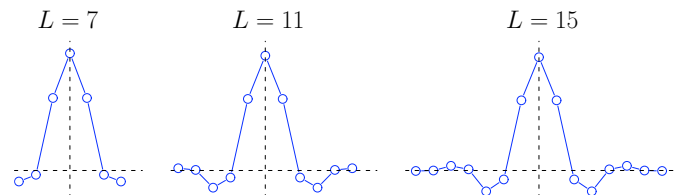
- for given  $L$ , there are multiple filters with the same  $\mathcal{G}_{(D)}(\cdot)$ , with these filters being distinguished by their phase functions  $\theta(\cdot)$ ; i.e., their transfer functions can be written as

$$G(f) \equiv \sum_{l=0}^{L-1} g_l e^{-i2\pi fl} = \mathcal{G}_{(D)}^{1/2}(f) e^{i\theta(f)}$$

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## Zero-Phase Filters

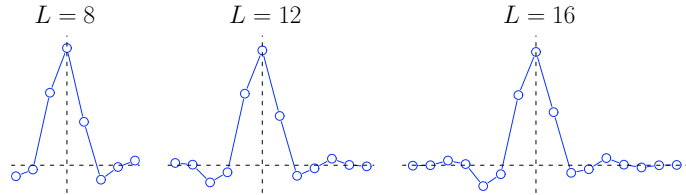
- Oppenheim and Lim (1981) note that filters with zero phase (i.e.,  $\theta(f) = 0$  for all  $f$ ) are important for eliminating distortions in filtered signals (particularly in images)
- zero-phase filters also facilitate aligning filter output with input
- conventional zero-phase filters  $\{a_l\}$  must be of *odd* length, say  $L = 2M + 1$ , and take the form  $a_{-l} = a_l$  for  $l = -M, \dots, M$
- three examples of zero-phase filters



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### 'Least Asymmetric' Scaling Filters (Symlets)

- in recognition of importance of zero-phase filters, Daubechies (1992) uses spectral factorization to obtain filters of widths  $L = 8, 10, 12, \dots$  closest to having zero phase (after a reindexing)
- three members of her class of 'least asymmetric' scaling filters



- cannot achieve filters with *exact* zero phase under her scheme because  $L$  must be even

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### Zero-Phase Wavelet (Zephlet) Transform: I

- possible to construct orthonormal DWT based on filters whose squared gain functions are consistent with those of Daubechies, but with *exact* zero phase, as following theorem states

- let  $\mathcal{G}(\cdot)$  and  $\mathcal{H}(\cdot)$  be squared gain functions satisfying

$$\mathcal{G}\left(\frac{k}{N}\right) + \mathcal{G}\left(\frac{k}{N} + \frac{1}{2}\right) = 2 \quad \text{and} \quad \mathcal{H}\left(\frac{k}{N}\right) + \mathcal{G}\left(\frac{k}{N}\right) = 2 \quad \text{for all } \frac{k}{N}$$

- let  $\{\bar{g}_l\}$  &  $\{\bar{h}_l\}$  be inverse DFTs of the sequences  $\{\mathcal{G}^{1/2}(\frac{k}{N})\}$  &  $\{\mathcal{H}^{1/2}(\frac{k}{N})\}$ :

$$\bar{g}_l \equiv \frac{1}{N} \sum_{k=0}^{N-1} \mathcal{G}^{1/2}\left(\frac{k}{N}\right) e^{i2\pi kl/N}, \quad l = 0, 1, \dots, N-1,$$

with an analogous expression for  $\bar{h}_l$

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### Zero-Phase Wavelet (Zephlet) Transform: II

- define the  $\frac{N}{2} \times N$  matrices

$$\mathcal{D}_1 = \begin{bmatrix} \bar{h}_1 & \bar{h}_0 & \bar{h}_{N-1} & \bar{h}_{N-2} & \bar{h}_{N-3} & \cdots & \bar{h}_5 & \bar{h}_4 & \bar{h}_3 & \bar{h}_2 \\ \bar{h}_3 & \bar{h}_2 & \bar{h}_1 & \bar{h}_0 & \bar{h}_{N-1} & \cdots & \bar{h}_7 & \bar{h}_6 & \bar{h}_5 & \bar{h}_4 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots & \vdots & \vdots \\ \bar{h}_{N-1} & \bar{h}_{N-2} & \bar{h}_{N-3} & \bar{h}_{N-4} & \bar{h}_{N-5} & \cdots & \bar{h}_3 & \bar{h}_2 & \bar{h}_1 & \bar{h}_0 \end{bmatrix}$$

and

$$\mathcal{C}_1 = \begin{bmatrix} \bar{g}_0 & \bar{g}_{N-1} & \bar{g}_{N-2} & \bar{g}_{N-3} & \bar{g}_{N-4} & \cdots & \bar{g}_4 & \bar{g}_3 & \bar{g}_2 & \bar{g}_1 \\ \bar{g}_2 & \bar{g}_1 & \bar{g}_0 & \bar{g}_{N-1} & \bar{g}_{N-2} & \cdots & \bar{g}_6 & \bar{g}_5 & \bar{g}_4 & \bar{g}_3 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \cdots & \vdots & \vdots & \vdots & \vdots \\ \bar{g}_{N-2} & \bar{g}_{N-3} & \bar{g}_{N-4} & \bar{g}_{N-5} & \bar{g}_{N-6} & \cdots & \bar{g}_2 & \bar{g}_1 & \bar{g}_0 & \bar{g}_{N-1} \end{bmatrix}$$

(note that, while  $\mathcal{D}_1$  has a form analogous to  $\mathcal{W}_1$  &  $\mathcal{V}_1$ , rows of  $\mathcal{C}_1$  are circularly shifted to the left by one)

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### Zero-Phase Wavelet (Zephlet) Transform: III

- then the  $N \times N$  matrix formed by stacking  $\mathcal{D}_1$  on top of  $\mathcal{C}_1$  is a real-valued orthonormal matrix; i.e.,

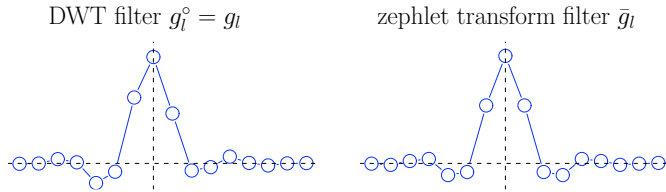
$$\mathcal{D} \equiv \begin{bmatrix} \mathcal{D}_1 \\ \mathcal{C}_1 \end{bmatrix} \quad \text{is such that} \quad \mathcal{D}^T \mathcal{D} = I_N$$

- moreover, the zero-phase circular filters  $\{\bar{h}_l\}$  and  $\{\bar{g}_l\}$  are related by  $\bar{g}_l = (-1)^l \bar{h}_l$  (note that this is in contrast to what holds for DWT filters, namely,  $g_l = (-1)^{l+1} h_{L-1-l}$ )
- proof of above theorem is similar in spirit to proof that  $\mathcal{W}$  is orthonormal, but details differ
- algorithms for computing DWT and zephlet transform are, respectively,  $\mathcal{O}(N)$  and  $\mathcal{O}(N \cdot \log_2(N))$

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### Zero-Phase Wavelet (Zephlet) Transform: IV

- for case  $N = L = 16$ , let's compare values in rows of  $\mathcal{V}_1$  based on Daubechies' least asymmetric filter and corresponding  $\mathcal{C}_1$  (after alignments for easier comparison)

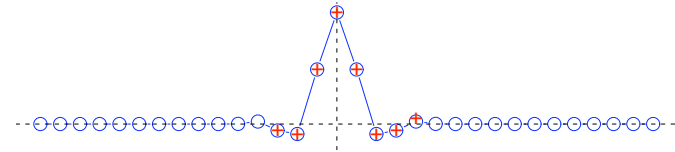


- for any  $N$  and  $L$ , squared magnitudes of DFTs of  $\{g_l^o\}$  &  $\{\bar{g}_l\}$  at  $f_k = k/N$  are exactly the same, but phase functions differ, with that for  $\{\bar{g}_l\}$  given by  $\theta(f_k) = 0$

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### Zero-Phase Wavelet (Zephlet) Transform: V

- for fixed  $L \geq 8$ , values in rows of zephlet transform change as  $N$  increases (DWT rows just add more 0's for all  $N \geq L$ )
- consider zephlet transform based on least asymmetric filter for  $L = 8$  and cases  $N = 8$  (pluses) and  $N = 32$  (circles)



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### Zero-Phase Wavelet (Zephlet) Transform: VI

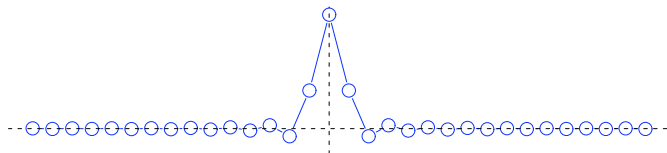
- can work out expression for elements in zephlet transform explicitly in Haar case ( $L = 2$ ):

$$\bar{g}_l = \frac{\sqrt{2}}{N} \left[ 1 + (-1)^l S_{l,+} + (-1)^{l+1} S_{l,-} \right] \approx \frac{2(-1)^l \sqrt{2}}{\pi(1 - 4l^2)}$$

for large  $N$ , where

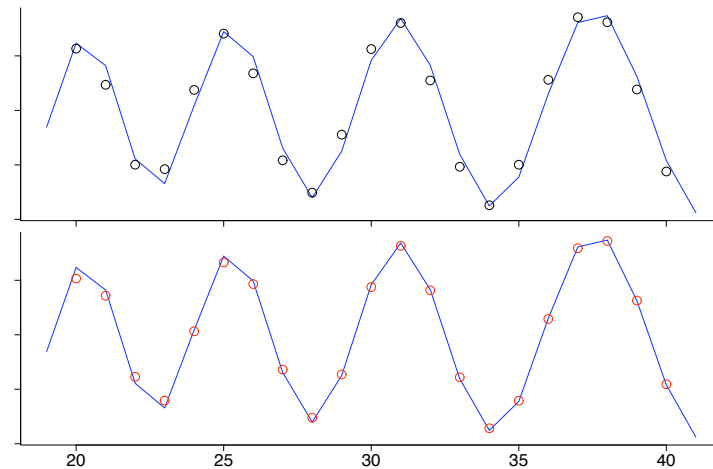
$$S_{l,\pm} \equiv \sin((2l \pm 1)\pi \frac{M-1}{4M}) \frac{\sin(\pi \frac{2l \pm 1}{4})}{\sin(\pi \frac{2l \pm 1}{4M})}$$

- Haar-based  $\{\bar{g}_l\}$  for  $N = 32$ :



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### Comparison of Outputs from LA(8) & Zephlet Scaling Filters (Input is Doppler Signal)



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## Concluding Remarks

- more work needed to elicit advantages/disadvantages of zephlet transform over usual DWT (in particular, for economic applications)
- can also formulate ‘maximal overlap’ version of zephlet transform (details in Percival, 2010)
- thanks to Ramo Gençay & conference organizers for opportunity to talk!
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