

INTRODUCTION

- Detection of signal discontinuities is an important problem in many signal processing tasks
- In applications such as MRI, sampled Fourier data is provided and we are required to locate the signal discontinuities
- Presence of Gibbs oscillations in a partial Fourier sum reconstruction impedes accurate detection of these jumps
- Here, we present the design and analysis of a detector based on the *concentration method* which computes the location, sign and magnitudes of jump discontinuities, given a finite number of noisy Fourier coefficients

THE CONCENTRATION METHOD

Let f be a 2π -periodic piecewise-smooth function with well-defined right and left hand limits. Its *jump function* is then defined as

$$[f](x) := f(x^+) - f(x^-)$$

To show that jump information is contained in Fourier data, consider a function with a single jump at $x = \gamma$. We can show that

$$\hat{f}(k) = [f](\gamma) \frac{e^{-ik\gamma}}{2\pi ik} + o\left(\frac{1}{k^2}\right)$$

Now consider a partial sum of the form

$$S_L[f](x) = \sum_{k=-L}^L \left(\frac{i\pi k}{L}\right) \hat{f}(k) e^{ikx}$$

Substituting for the Fourier coefficients, we would obtain

$$S_L[f](x) = [f](\gamma) \underbrace{\frac{1}{2L} \sum_{k=-L}^L e^{ik(x-\gamma)}}_{\text{scaled Dirichlet kernel}} + \sum_{k=-L}^L o\left(\frac{1}{k}\right) e^{ik(x-\gamma)}$$

i.e., the jump approximation “concentrates” at the singular support of f . More generally, we have

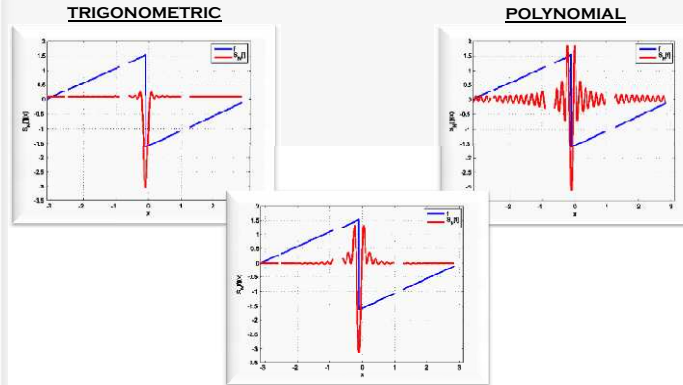
$$S_L^{\eta}[f](x) = i \sum_{k=-L}^L \hat{f}_k \operatorname{sgn}(k) \sigma\left(\frac{|k|}{L}\right) e^{ikx}$$

where $\sigma_{k,L}(\eta) = \sigma(|k|/L)$ are called concentration factors

The jump function approximation can be computed efficiently using a FFT.

CONCENTRATION FACTORS

- Several concentration factors are available for use, including



- Choice of concentration factor dictates tradeoff between spurious values away from the discontinuity and width of the “mainlobe” at the discontinuity

DETECTOR DESIGN

- We model noise as additive white Gaussian with zero mean

$$\hat{g}_k = \hat{f}_k + \hat{n}_k \quad \hat{n}_k \sim \mathcal{N}[0, \sigma^2]$$

- The concentration method is linear and the noise component does not bias the jump function approximation
- Choosing a length R signal vector,

$$M = (S_L^{\sigma_1}[f](x_1), \dots, S_L^{\sigma_R}[f](x_R))^T$$

The detection problem becomes,

$$\mathcal{H}_0: \mathbf{Y} = \mathbf{N} \sim \mathcal{N}[0, C_N]$$

$$\mathcal{H}_1: \mathbf{Y} = M + \mathbf{N} \sim \mathcal{N}[M, C_N]$$

- The Neyman-Pearson detector yields

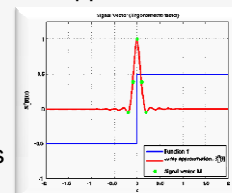
$$M^T C_N^{-1} \mathbf{Y} > \gamma$$

o γ is a threshold

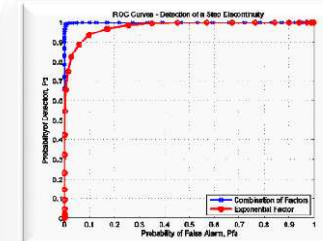
o Performance metric is $M^T C_N^{-1} M$

o Performance curve is described by

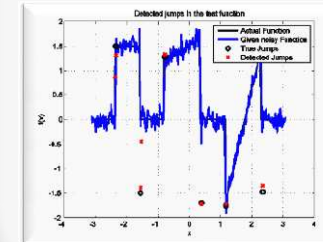
$$P_d = Q\left(Q^{-1}(P_{fa}) - \sqrt{M^T C_N^{-1} M}\right)$$



RESULTS



DETECTION OF A STEP DISCONTINUITY – ROC CURVES



DETECTION OF DISCONTINUITIES IN A TEST FUNCTION

- *Specifications:*

R = 3, L = 32, $\sigma^2 = 7.5$

• Red curve – Exponential concentration factor

• Blue curve – Combination of all factors

• SNR metrics: 9.028 dB (Red) and 13.434 dB (Blue)

- *Specifications:*

R = 3, L = 128, $\sigma^2 = 7.5$

• Trigonometric concentration factor used

CURRENT DIRECTIONS

- Multi-dimensional edge detection
- Sidelobe effect mitigation and false-alarm reduction
- Jump detection from non-uniform Fourier data

REFERENCES

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