

ABSTRACT

Individuals with nystagmus, a visual impairment characterized by involuntary, repetitive eye movements, perceive an unstable visual field, reducing their quality of life. The presented research contributes to the development of a novel aide for individuals with nystagmus – a mobile application (app), *StabilEyes*, for any smart device – to be used during daily activities. StabilEyes will use the front-facing camera to detect the user's unwanted periodic eye motion and translate real-time images captured by the back-facing camera on the screen, effectively stabilizing the visual field of the user. We have developed Discrete Period Quadrature (DPQ), a method for time-frequency analysis, for the purpose of tracking this periodic eye motion. DPQ operates on a discrete signal and produces complex values for a set of frequencies. The resulting magnitude and phase spectra have various features by which the fundamental signal period may be inferred. These features include maximum magnitude at the signal period, local minimum magnitudes at the signal period subharmonics, minimum variance of magnitude at the signal period, and minimum root-mean-squared (RMS) phase-change error at the signal period. We compared the capability of each of the four spectral features identified to infer the period and phase of a sinusoidal signal whose period was changed at various rates or that had various amounts of noise added. Some observations using nonsinusoidal signals (particularly ramp waves) were also made. Preliminary analyses suggest that finding the minimum RMS error between the observed and expected rate of phase-change is the optimal method for tracking a periodic signal with DPQ.

BACKGROUND

In the continuous-time domain, the Fourier Transform operates on a signal over a time window of $(-\infty, \infty)$ to determine the magnitude and phase of all constituent frequencies. The Discrete Fourier Transform (DFT) is the discrete-time equivalent.

Our method, DPQ, is founded upon the same concepts as the DFT and produces magnitude and phase spectra that are comparable to those of the DFT. However, unlike the DFT, DPQ utilizes a window whose size adapts to exactly one discrete period for each frequency considered. This optimizes DPQ for rapid and accurate tracking of a periodic signal.

For our purposes, DPQ must remain both fast and accurate in situations where (1) the fundamental frequency is changing and (2) noise is present.

PURPOSE STATEMENT

The presented work investigates the use of various spectral features produced by DPQ to infer the frequency and phase of a periodic signal, especially with a change in the signal period or in the presence of noise.



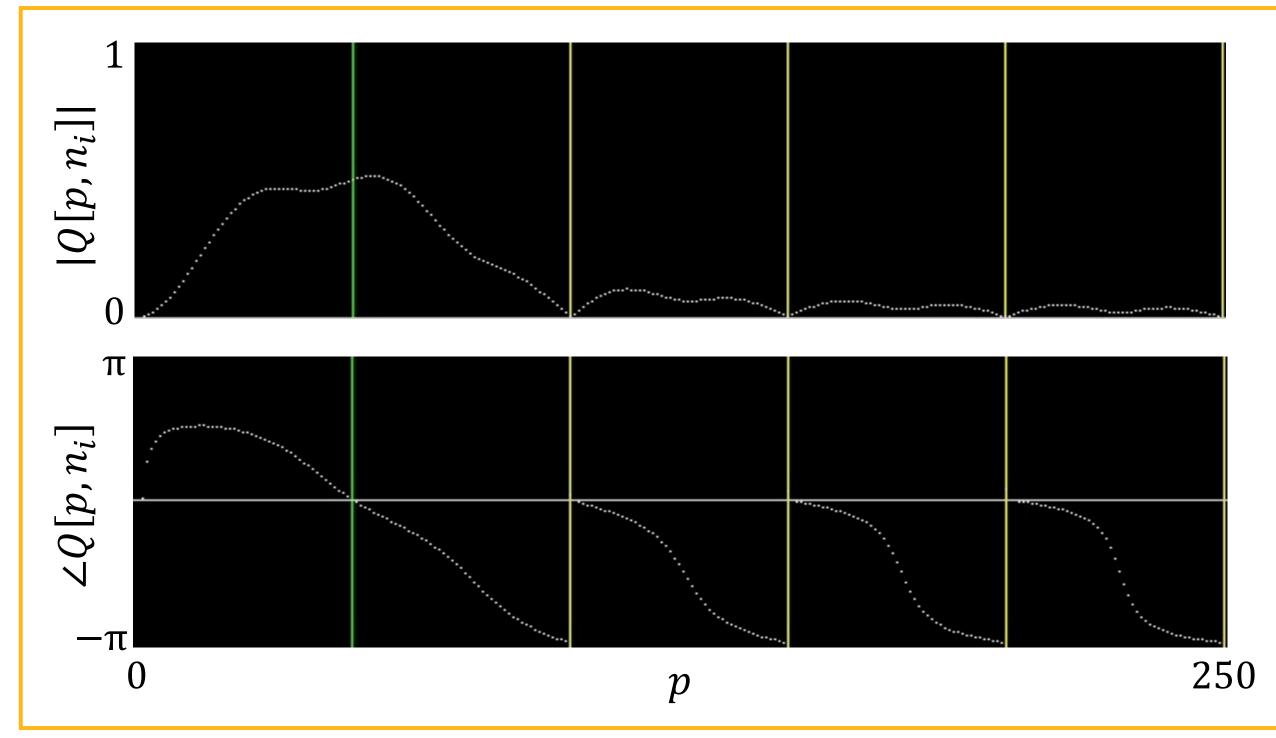
Discrete Period Quadrature for Time-Frequency Analysis of Nystagmus Eye Motion

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METHODS

$$Q[p,n] = \sum_{m=n}^{n-p+1} \frac{s[m]\left(\cos\left(\frac{2\pi(m-n)}{p}\right) + j\sin\left(\frac{2\pi(m-n)}{p}\right)\right)}{p}, 2 \le p \le P$$

for a signal with a period of 50 samples/cycle.



The green and yellow vertical lines denote the (true) signal period p_s and its subharmonics (integer multiples of p_s), respectively.

We identified the following four spectral features as useful in inferring p_s :

- <u>Peak</u>: |Q[p,n]| has a global maximum at $p \cong p_s$
- <u>Subharmonic</u>: |Q[p,n]| has local minima at $p \cong kp_s$, k = 2, 3, 4, ...
- <u>Delta Phase</u>: RMS error of $\left[\angle Q[p,n] \angle Q[p,n-1] \right] \frac{2\pi}{n}$ has a global minimum at $p \cong p_s$

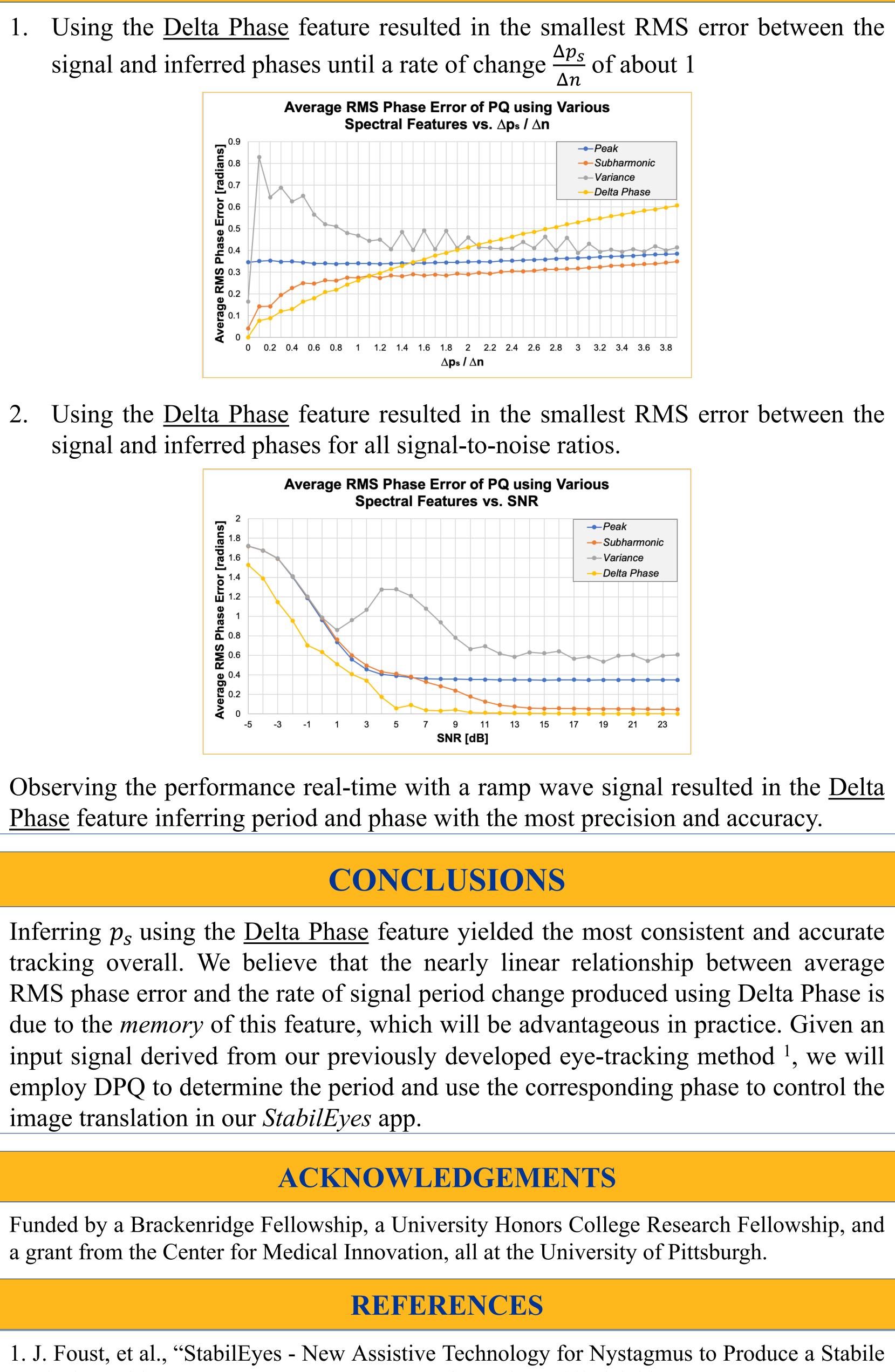
We conducted two experiments using a sinusoidal signal with (1) a change in signal period and (2) Gaussian noise, in which the phase corresponding to each inferred period was compared to the true phase of the signal at each index.

We also observed the real-time performance of each feature using a **non**sinusoidal (ramp) waveform as the input signal.

DPQ operates on a discrete signal, s[n], yielding complex values for a set of frequencies. Each frequency ω radians/sample has a corresponding integer period p samples, such that $\omega = \frac{2\pi}{n}$. We denote DPQ as Q[p,n] and define it

where the sum operates on p samples preceding and including the sample at index n. For each p between 2 and a maximum P, this yields the quadrature covariance at each discrete frequency ω as a complex number. One example of the resulting magnitude and |Q[p,n]| phase $\angle Q[p,n]$ spectra is shown below

<u>Variance</u>: VAR(|Q[p,n]|) has local minima at $p \cong kp_s$, k = 1, 2, 3, ...





RESULTS & DISCUSSION

Real-Time Video Image" at BMES 2019 Annual Meeting, Philadelphia, PA, 2019.