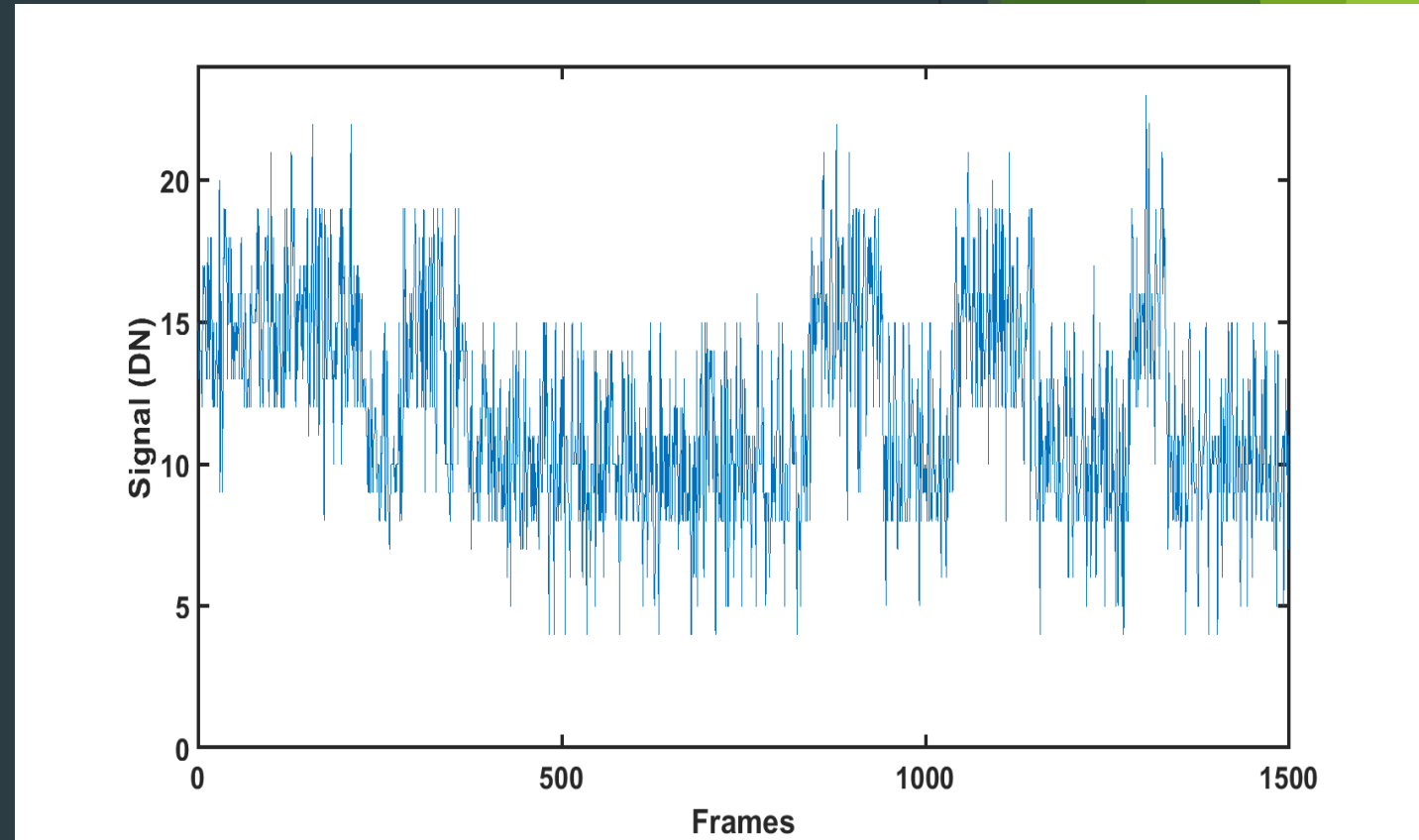


# Using Wavelets to Analyze Random Telegraph Signal Noise

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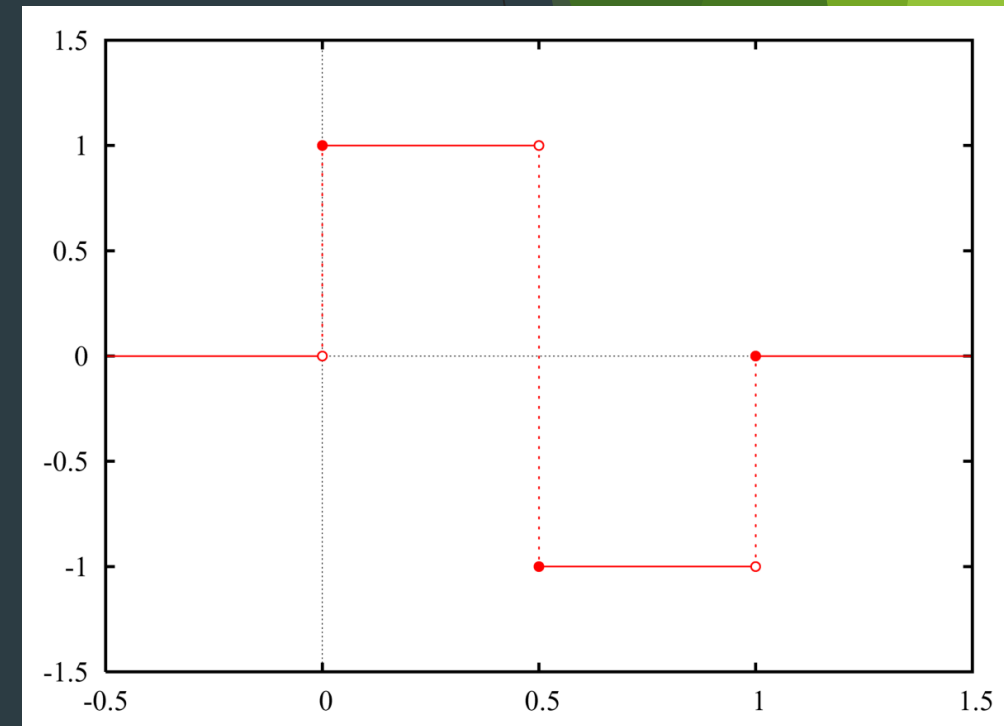
# RTS Noise - Overview

- ▶ Observed in CCD and CMOS architectures
- ▶ Defined by discrete changes in signal level (blinking pixels)
- ▶ Stochastic process with Poisson distributed state lifetimes
- ▶ Characterized by similar lifetimes



# The Haar Wavelet & Discrete Wavelet Transform

- ▶ An orthonormal basis set developed by Alfred Haar in 1909
- ▶ Left largely in obscurity until DeBauchies pioneering work constructing and using wavelets for digital signal processing and analysis
- ▶ DWT Useful for edge detection applications
- ▶ Acts like a microscope, signals analyzed at a variety of scales



# Experimental Parameters for radiation effects in Si image sensors

- ▶ COTS Omnivision OV5647
- ▶ Raw frames taken using a Raspberry Pi 3
- ▶ Five sensors irradiated with a continuum of high energy  $\gamma$  and x-rays (peak -  $2MeV$ )
- ▶ 1500 frames taken at 0.05 frames/s
  - ▶ ~ 8.3 hours total measurement time
- ▶ Frames taken in dark at 23°C

Absorbed Ionizing Dose (rad - Si)
500
2,500
5,000
10,000
25,000



# Haar Wavelet Analysis - DWT

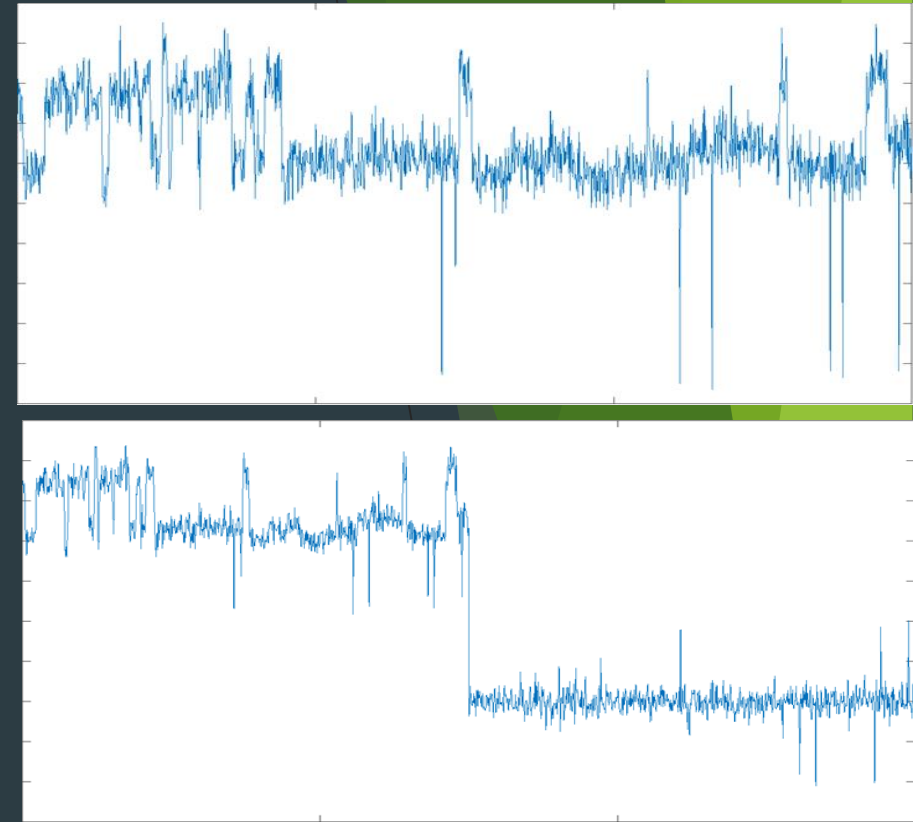
- ▶ Consider a digital signal  $\mathbf{f} = (f_1, f_2, f_3, \dots, f_N)$
- ▶ The discrete wavelet transform (DWT) breaks  $\mathbf{f}$  into two 'daughter' series of length  $N/2$

- ▶ Trend Series Members

- ▶  $a_m = \frac{f_{2m-1} + f_{2m}}{\sqrt{2}} \quad 1 < m \leq N/2$

- ▶ Details Series Members

- ▶  $d_m = \frac{f_{2m-1} - f_{2m}}{\sqrt{2}} \quad 1 < m \leq N/2$



# Haar Wavelet Analysis - DWT (cont.)

- ▶ The DWT is similar to a microscope because it is repeatable
  - ▶ The trend series is treated as the new 'mother' signal
- ▶ Each time a subsequent transform is performed the 'daughter' series are of half size
  - ▶ The new 'daughter' series represent twice as many values from the original signal

## Wavelet Operators - Trends Series

$$\blacktriangleright \mathbf{V}_1^1 = \frac{1}{\sqrt{2}} (1, 1, 0, 0, 0, \dots); \quad \mathbf{V}_2^1 = \frac{1}{\sqrt{2}} (0, 0, 1, 1, 0, \dots)$$

$$a_1 = \frac{f_1 + f_2}{\sqrt{2}} = \mathbf{f} \cdot \mathbf{V}_1^1; \quad a_2 = \frac{f_3 + f_4}{\sqrt{2}} = \mathbf{f} \cdot \mathbf{V}_2^1$$

$$a_m = \frac{f_{2m-1} + f_{2m}}{\sqrt{2}} = \mathbf{f} \cdot \mathbf{V}_m^1$$

## Wavelet Operators - Details Series

$$\blacktriangleright \mathbf{W}_1^1 = \frac{1}{\sqrt{2}} (1, -1, 0, 0, 0, \dots), \quad \mathbf{W}_2^1 = \frac{1}{\sqrt{2}} (0, 0, 1, -1, 0, \dots)$$

$$d_1 = \frac{f_1 - f_2}{\sqrt{2}} = \mathbf{f} \cdot \mathbf{W}_1^1; \quad d_2 = \frac{f_3 - f_4}{\sqrt{2}} = \mathbf{f} \cdot \mathbf{W}_2^1$$

$$d_m = \frac{f_{2m-1} - f_{2m}}{\sqrt{2}} = \mathbf{f} \cdot \mathbf{W}_m^1$$



# The Inverse Transform

$$\mathbf{f} = \left( \frac{a_1 + d_1}{\sqrt{2}}, \frac{a_1 - d_1}{\sqrt{2}}, \dots, \frac{\frac{a_N + d_N}{2}}{\sqrt{2}}, \frac{\frac{a_N - d_N}{2}}{\sqrt{2}} \right)$$

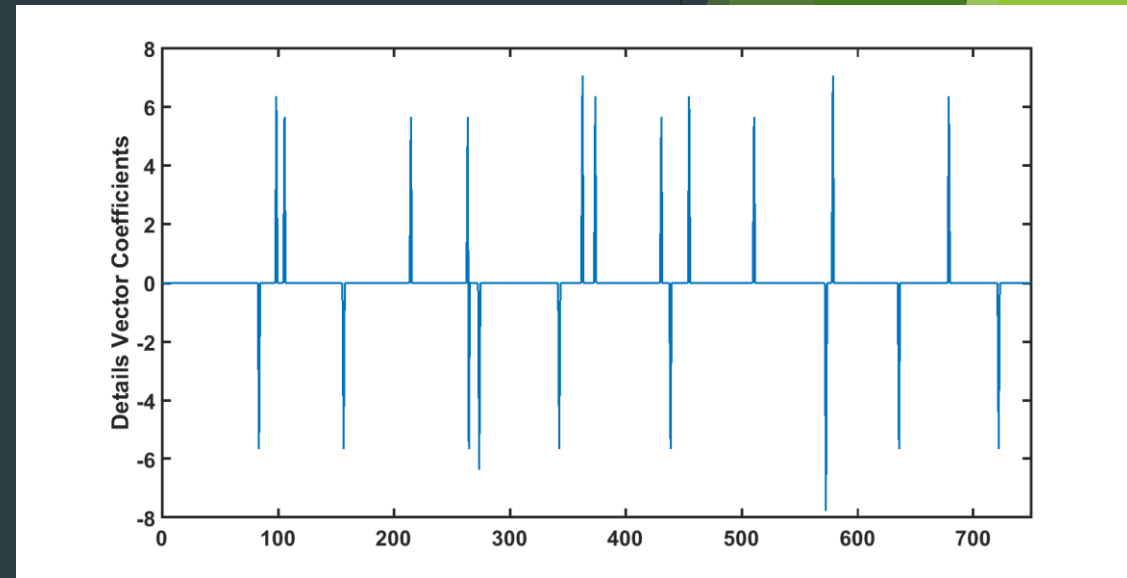
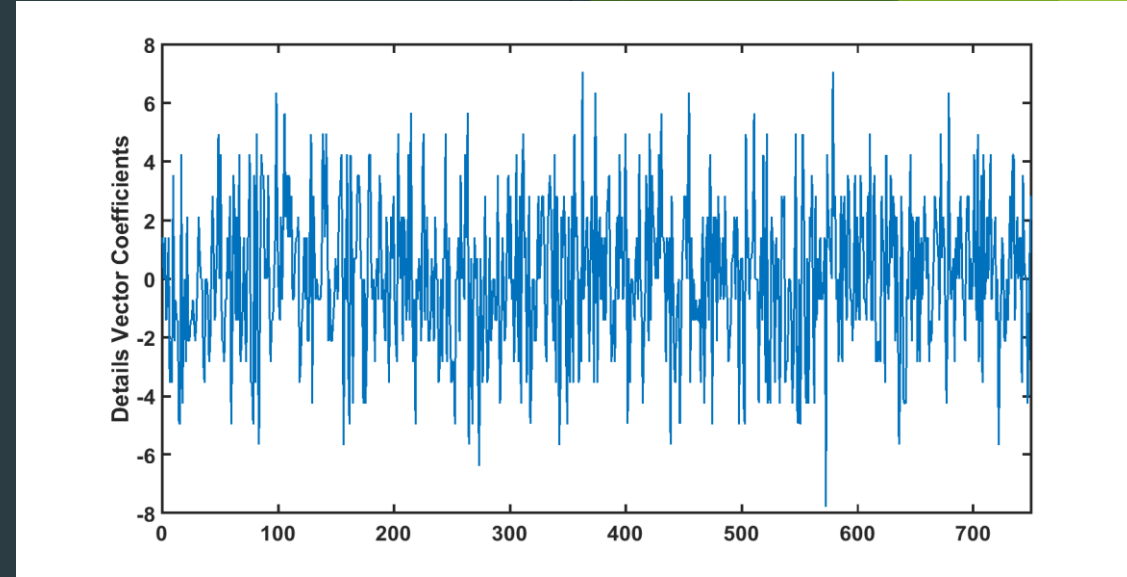
# DWT - Denoising

- ▶ White noise is suppressed by thresholding the details series
  - ▶ Similar to a high-pass/low-pass filter
  - ▶ Based on magnitude rather than frequency
- ▶ The threshold is statistically derived

$$T = \hat{\sigma} \sqrt{2 \log(n)}$$

- ▶  $T$  is the universal threshold derived by Donoho and Johnstone<sup>†</sup>
- ▶ Values below the threshold are set to zero

<sup>†</sup>G. P. Nason, “Choice of the threshold parameter in wavelet function estimation,” *Wavelets and statistics*, vol. 103, pp. 261–280, 1995.



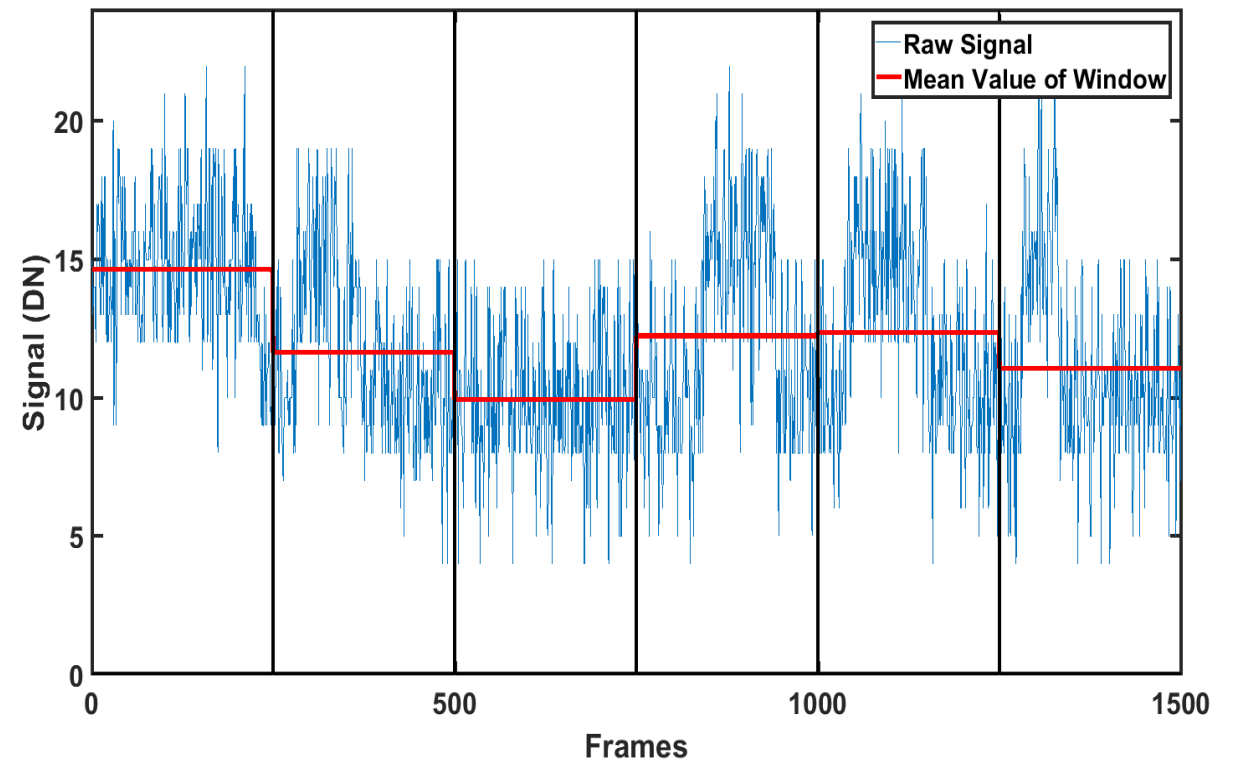
# Approximation Signal Construction

- ▶ The goal is to remove all noise from the signal except for the RTS transitions
- ▶ The DWT is performed seven times and thresholded at every iteration
- ▶ The chosen threshold is designed be highly discriminatory to prevent false positive detections
- ▶ A temporal screen is implemented to remove contributions from single events like cosmic rays

# Approximation Signal Construction

## Stage 1: Window Comparison

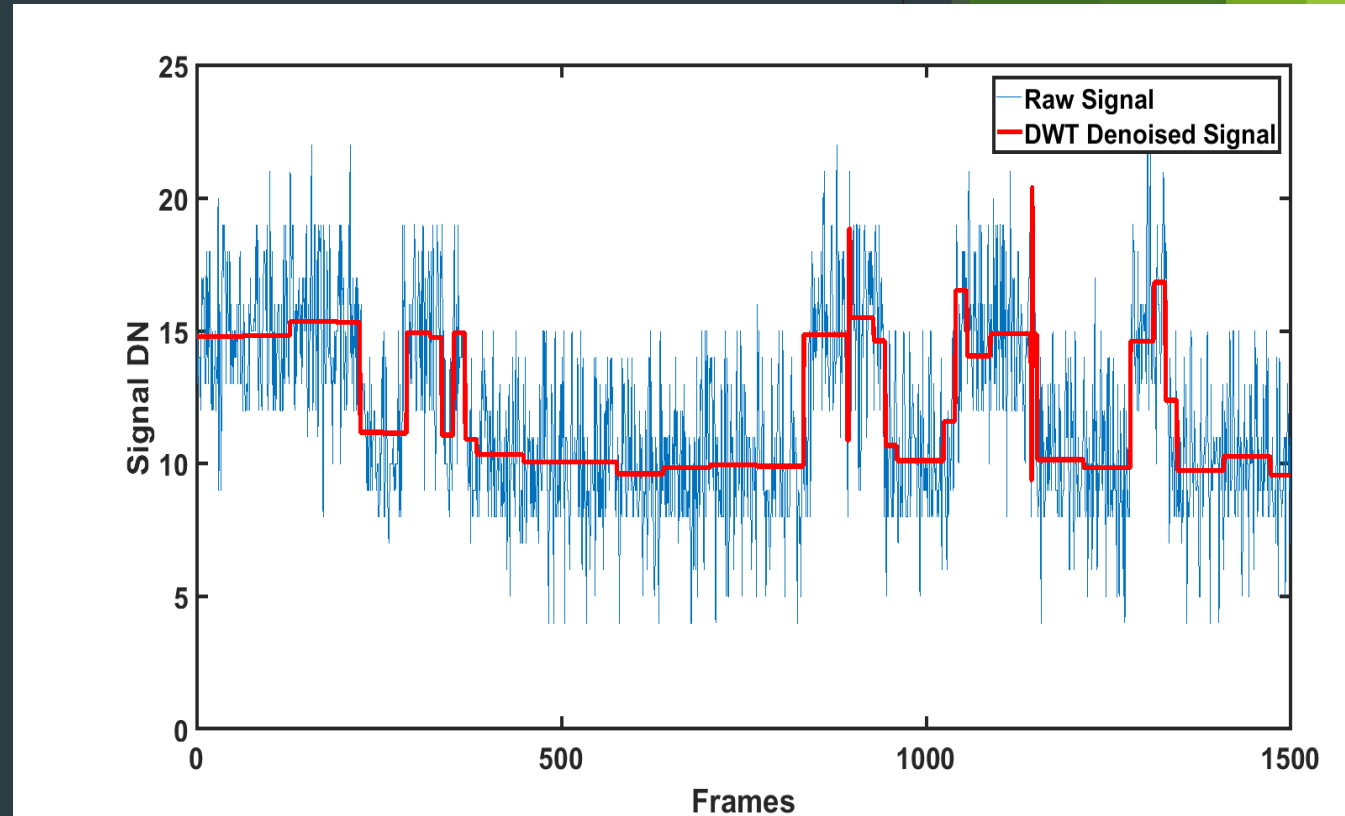
- ▶ Cut signal into six windows
- ▶ Compare the mean of a window to the previous two
- ▶ If any difference is greater than the standard deviation of the noise  $\sigma_r$  the signal progresses as an RTS candidate
- ▶ Crude, but highly effective and discriminatory



# Approximation Signal Construction

## Stage 2: DWT Denoising

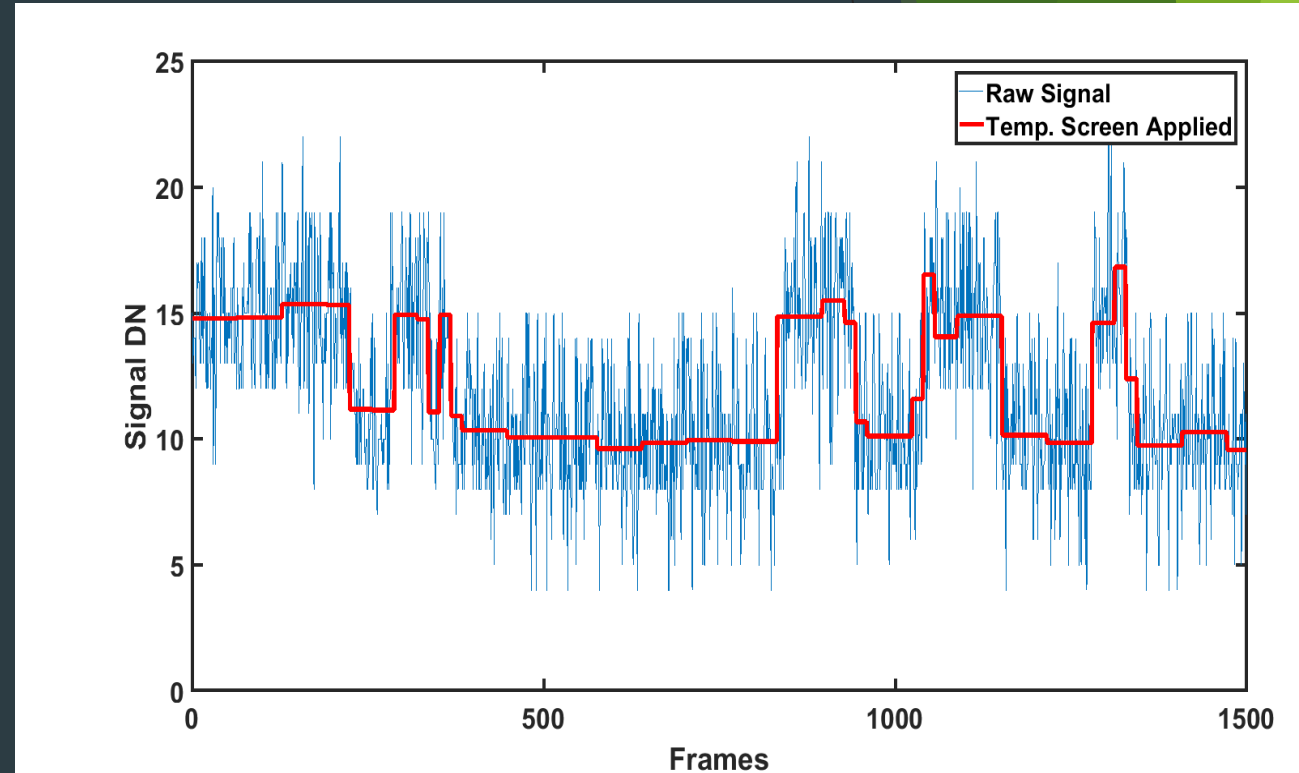
- ▶ The signal is run through the DWT denoising method as described previously
- ▶ The white noise is greatly reduced, but a few transients remain



# Approximation Signal Construction

## Stage 3: Temporal Screen

- ▶ To remove transients, a simple running comparison is implemented to verify the stability of a transition
- ▶ When a change in magnitude happens at frame  $k$ , its value is compared to the next  $l$  frames where  $l = 10$
- ▶ If the value is unchanged the transition is considered stable and left alone
- ▶ If the value changes is considered a transient, and is changed to the value at frame  $k - 1$

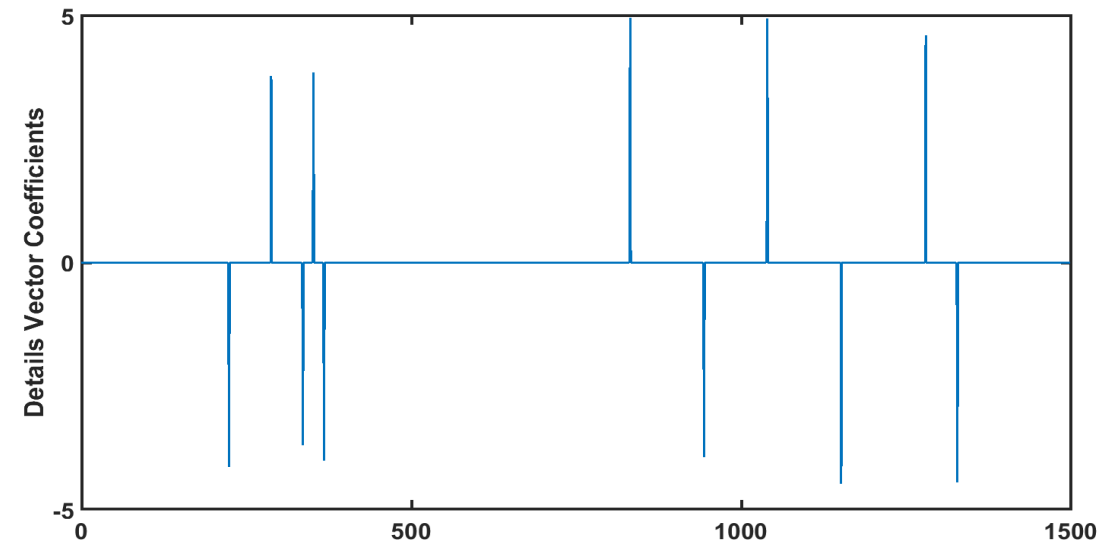
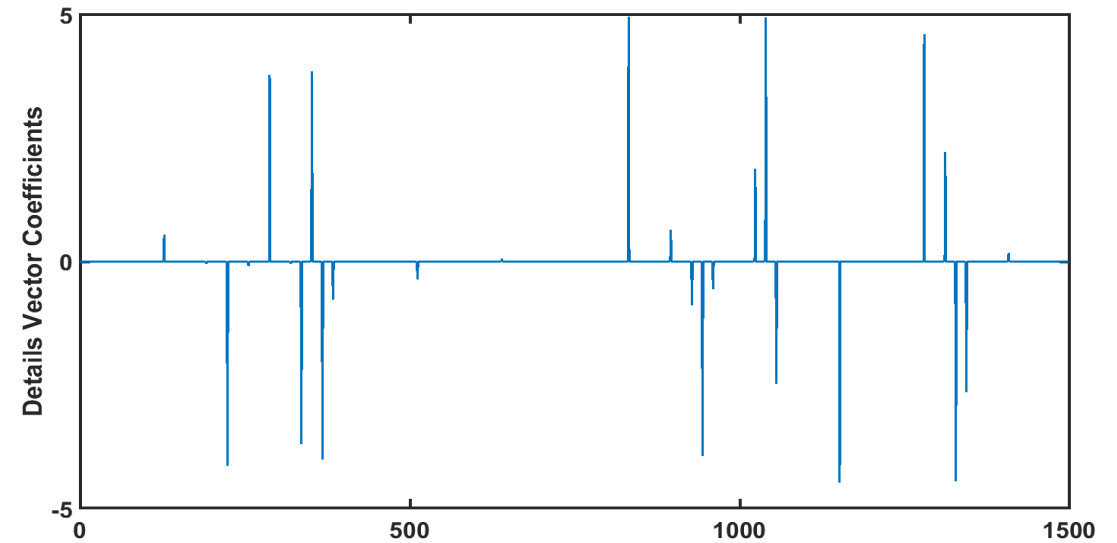


# Approximation Signal Construction

## Stage 4: A Second Thresholding

- ▶ Nearly all of the white noise is removed, but a few small changes remain
- ▶ A new details series is creating by subtracting each frame value by the previous frame value
- ▶ The new details series  $s$  is of  $N - 1$  where  $N$  is the size of the original signal
- ▶ Because the noise is already suppressed, the threshold need not be so discriminatory, as such

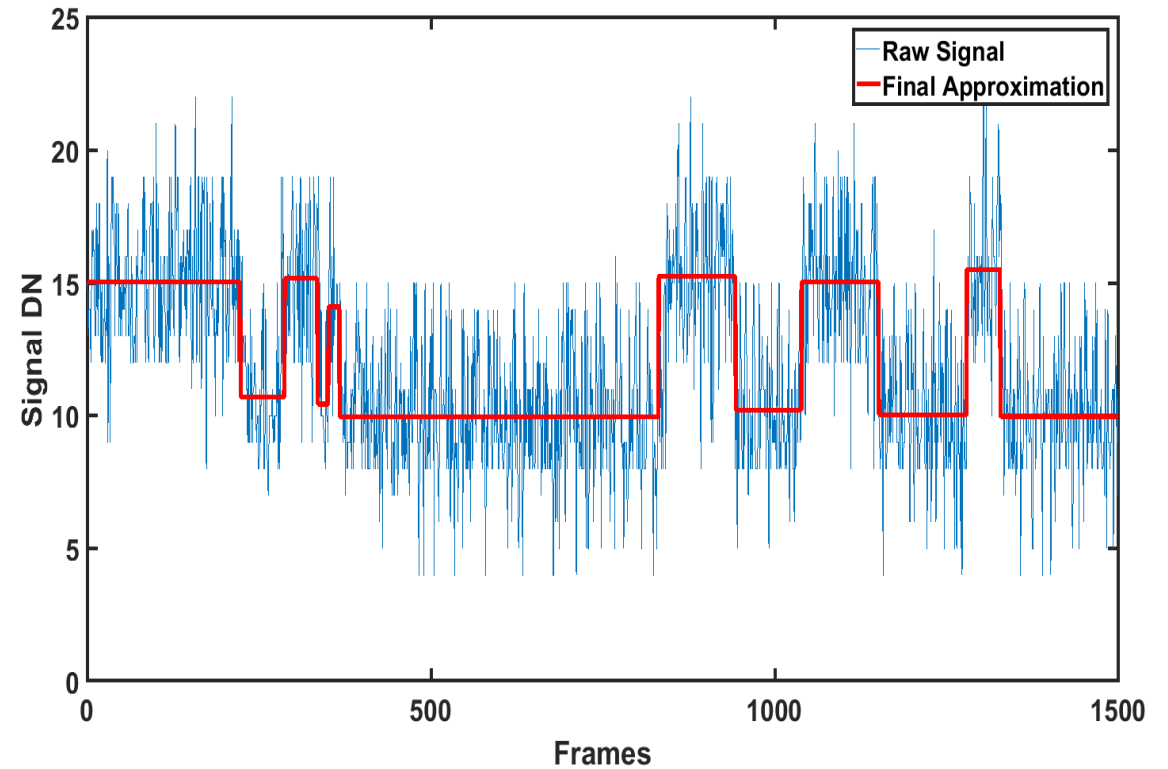
$$T_s = S_{MAX} * u_{0.75}$$



# Approximation Signal Construction

## Stage 5: Final Reconstruction

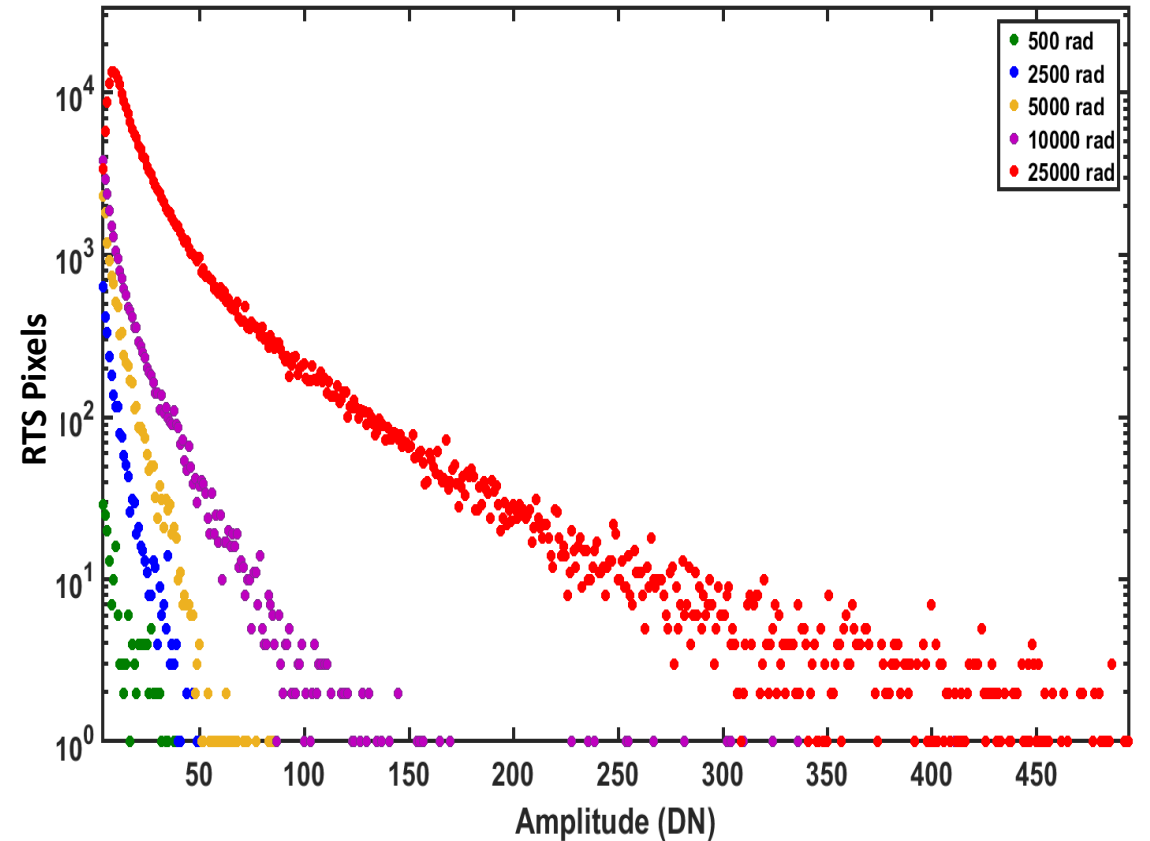
- ▶ The final approximation is created by applying the mean of the original signal to the segments between the remaining non-zero values of the latest details series
- ▶ With the new approximation complete the RTS amplitudes and time constants can be gathered easily for analysis





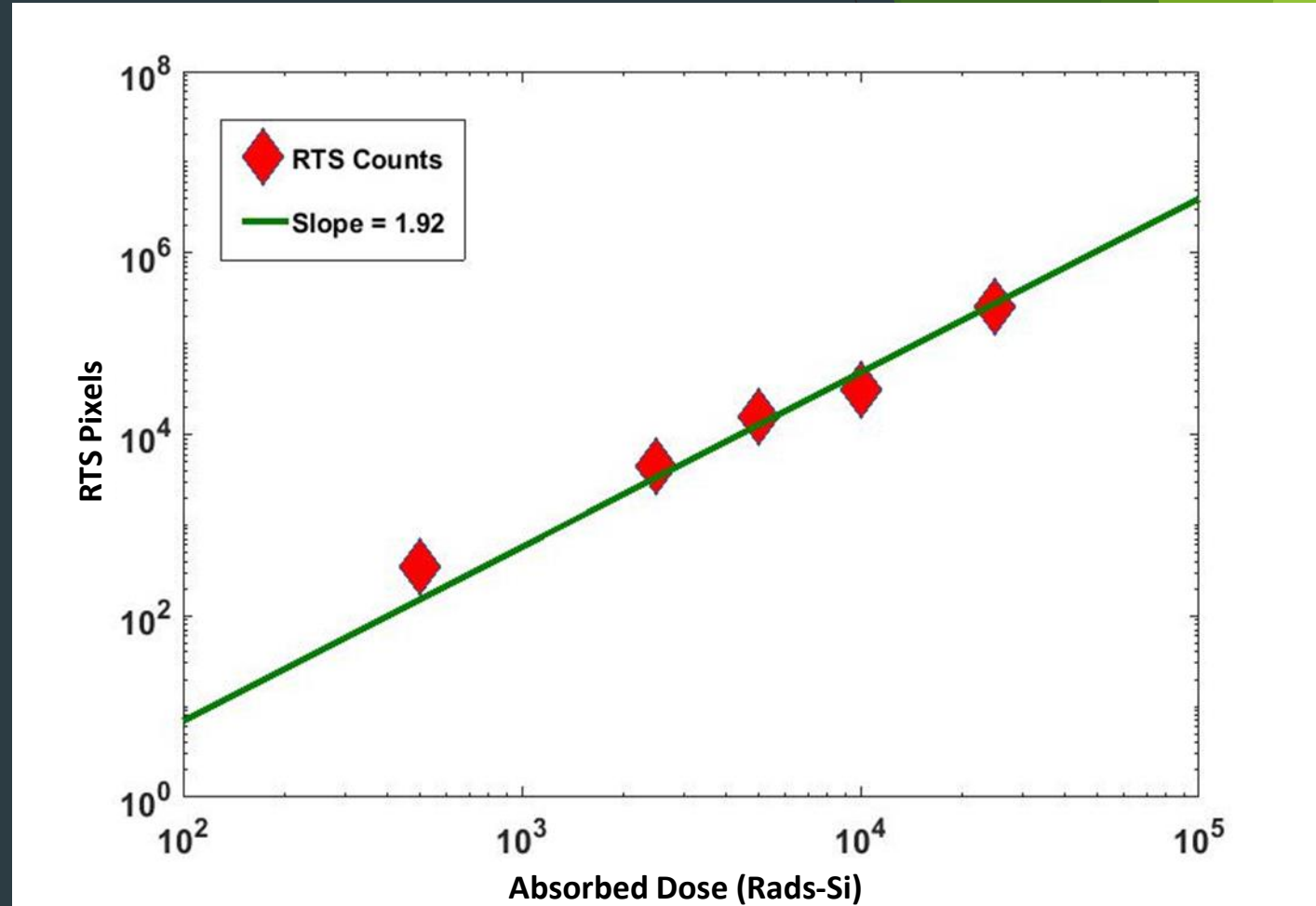
# Results - Maximum Amplitudes

- ▶ Similar shape of curves indicates that higher doses increase the likelihood of creating an RTS center, but the amplitude probability for a center is set
- ▶ No correlation seen between RTS amplitude and time constants



# Results - Second-Order Defect Generation

- ▶ The number of RTS centers increases ~quadratically with absorbed dose
- ▶ This indicates that the particular defect responsible for this RTS noise is of second-order



# Thank You

## Acknowledgements

- ▶ Dr. Richard Crilly from Oregon Health and Science University for help with sensor irradiation
- ▶ My labmates: Justin Dunlap, Bahar Ajdari, Joe Niederriter, Paul DeStefano, and Denis Heidtmann



# Results - State Lifetimes

- ▶ Lifetimes are calculated by averaging the time spent in the high or low states
- ▶ Both high and low states display an exponential distribution
- ▶ The low state time constant distribution is slightly flatter than the high state, indicating that the low state is the more stable of the two

