

Dust Devil Heartbeat Detection on Infrasound Sensors

Elizabeth M. Berg¹, Daniel C. Bowman¹, Elizabeth A. Silber¹, Léo Martire², Siddharth Krishnamoorthy², Attila Komjathy²

¹Sandia National Laboratories, Albuquerque, New Mexico

²NASA Jet Propulsion Laboratory, California Institute of Technology, Pasadena, California

Abstract

Particle-loaded convective vortices, or dust devils, are coherent, columnar shaped, fast rotating, dust-laden vortices at least one meter high and lasting for at least 10 seconds. These **vortices are created from heating of near-surface air by insolation and so are subject to boundary layer processes**. Dust devils consist of reduced pressure at their centers and are observed moving across pressure sensors as a pressure dip of varying duration (full-width at half maximum, FWHM) and amplitude, a function of dust devil radius and magnitude. These result in large, prominent signals on low frequency sound, or infrasound, stations.

However, **on raw infrasound data, this signal appears as a distinctly heartbeat shape due to convolution of the dust devil pressure dip and the instrument response**. We create a series of heartbeat templates spanning a variety of expected dust devil signals and, via **template-matching cross-correlation**, identify and characterize dust devils within infrasound time-series data. We statistically quantify the limitations and potential applications of this method to recorded infrasound data from multiple sensors in the Mojave desert in southern Nevada.

In addition, we develop an **empirical background noise model** for the Mojave desert infrasound sensors and test the capability of **wavelet transform-based methods** to differentiate dust devil signatures from background atmospheric variability.

Not only is this of scientific interest in characterization of the planetary boundary layer, but it also has implications for the risk to life and property posed by these vortices. Furthermore, automatic dust devil detection may improve understanding of these signatures against instrument noise, and applicable to understanding the occurrence and variability of dust devils on Mars, which is answerable to several high-priority goals set forth by the Mars Exploration Program Analysis Group (MEPAG).

Correlation Heartbeat Detector

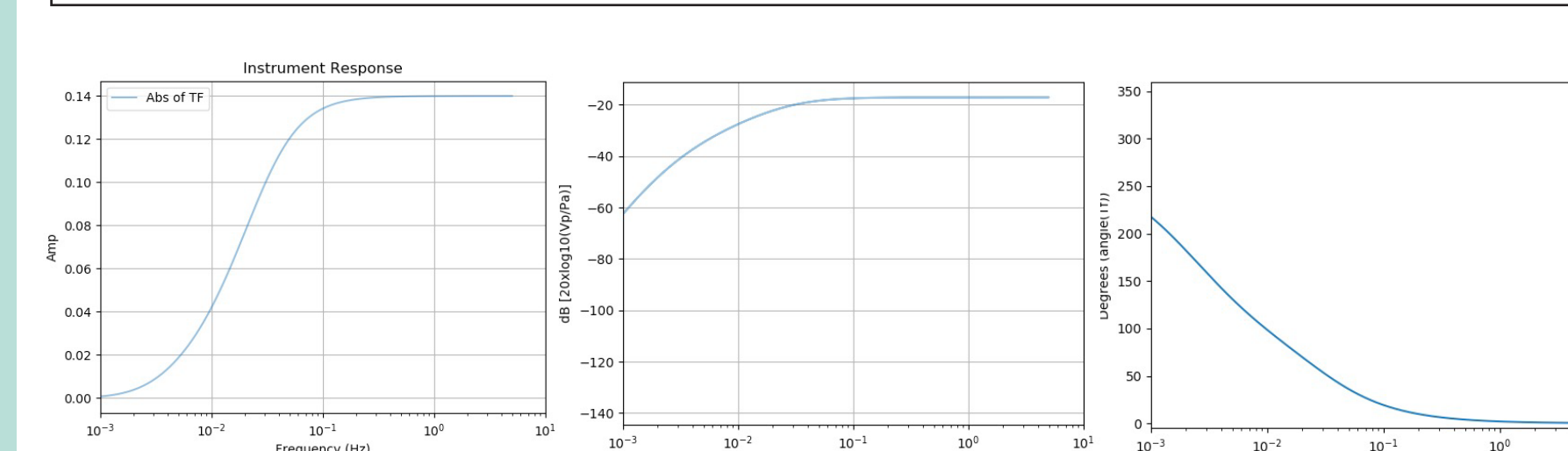
Goal: Build a detector of the characteristic vortex heartbeat signal observed on Hyperion pressure sensors
Steps:

1. Build templates of synthetic vortex pressure signals as expected for Hyperion pressure sensors
2. Create algorithm and methodology to efficiently calculate the correlation coefficient between templates and data streams
3. Test robustness of correlation method on synthetics
4. Apply to a 6 hour period of recorded data across multiple sensors

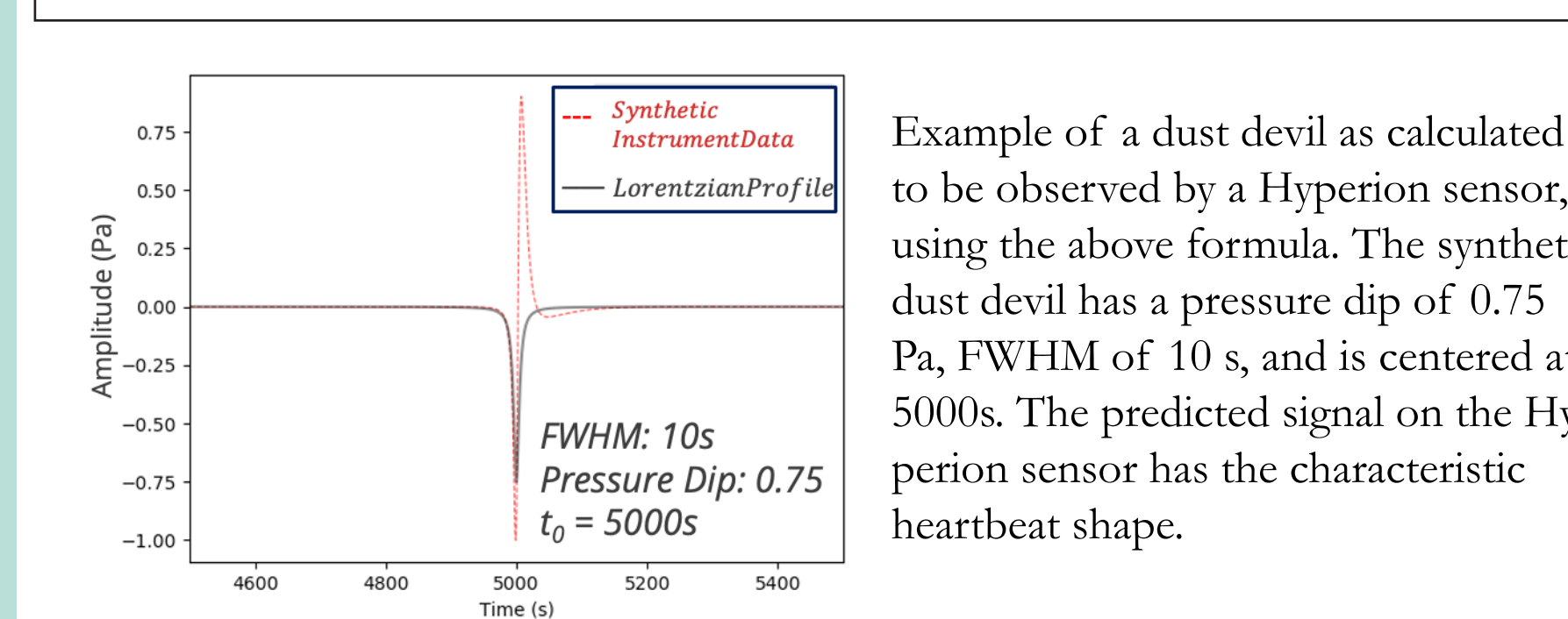
$$C_i(t) = \frac{x_i(t_0) \cdot y_i(t)}{\sqrt{x_i(t_0) \cdot x_i(t_0)} \sqrt{y_i(t) \cdot y_i(t)}} \quad \text{Gibbons \& Ringdal (2006)}$$

1. Build templates of expected vortex pressure signals

$$H(f) = \frac{\text{Hyperion Instrument Transfer Function}}{(f - 0.001483i)(f - 0.003387i)(f - .02949i)} \quad \text{Merchant (2015)}$$



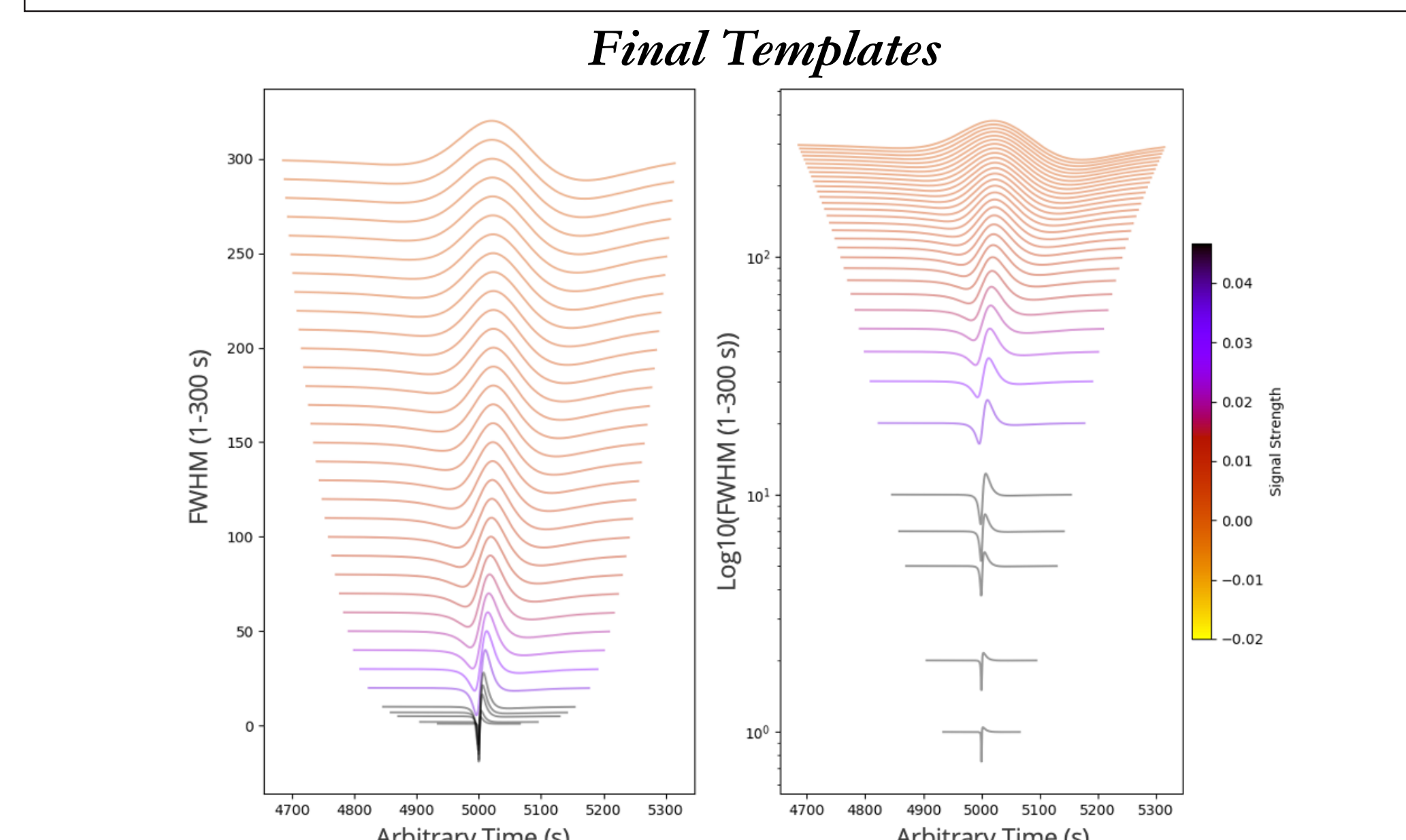
$$\text{Synthetic Dust Devil Vortex for Hyperion Sensors} \\ \text{SyntheticInstrumentData}(f) = \text{LorentzianProfile}(f) \cdot H(f)$$



Example of a dust devil as calculated to be observed by a Hyperion sensor, using the above formula. The synthetic dust devil has a pressure dip of 0.75 Pa, FWHM of 10 s, and is centered at 5000s. The predicted signal on the Hyperion sensor has the characteristic heartbeat shape.

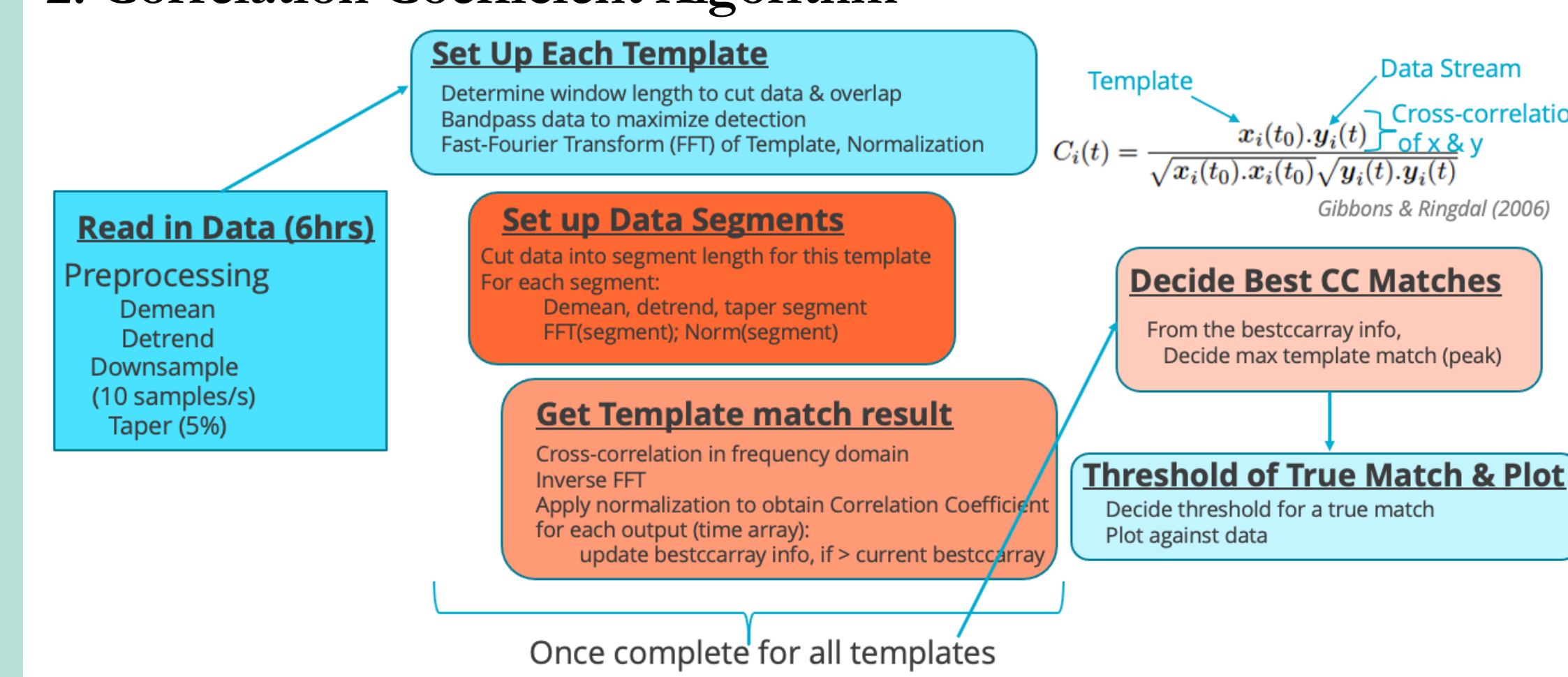
$$\text{LorentzianProfile}(t) = \frac{\text{PressureDip}}{1 + (2(t - t_0)/\text{FWHM})^2 + B} \quad \text{Jackson \& Lorenz (2015)}$$

FWHM: Full-Width Half Maximum; t_0 = Time of max dip

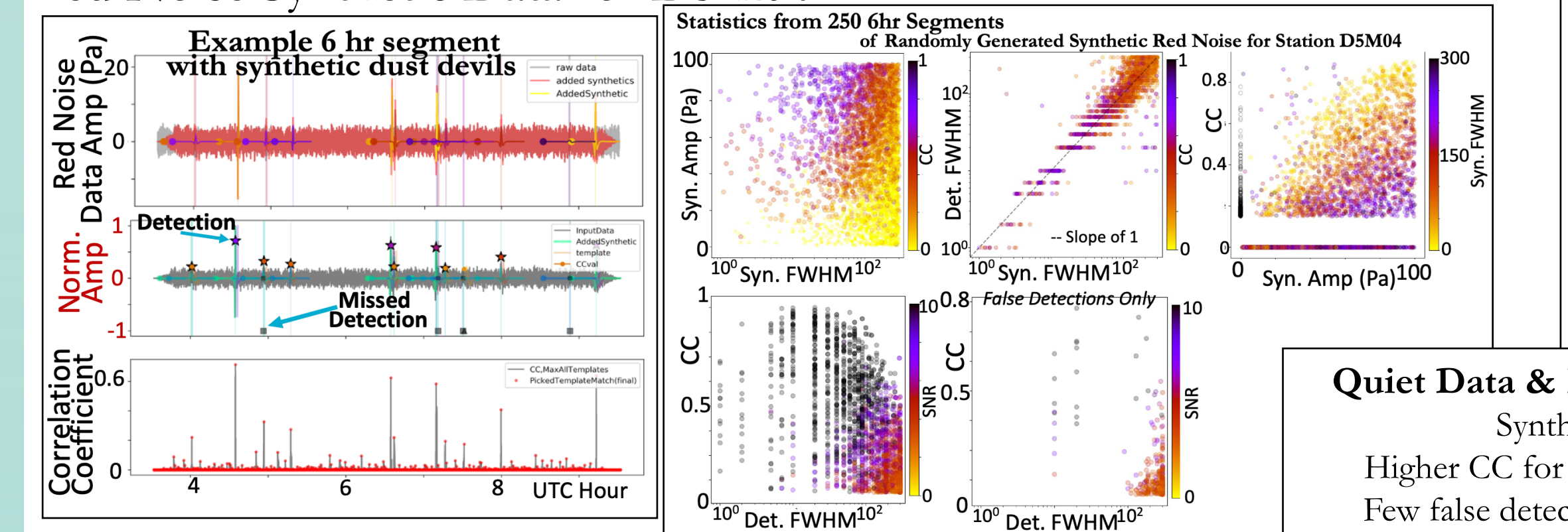


As the correlation coefficient is sensitive to the phase (not amplitude) of the signal, our templates are created via a range of FWHM, which alters the shape. We note that the signal strength is much lower for FWHM > approximately 100 s.

2. Correlation Coefficient Algorithm

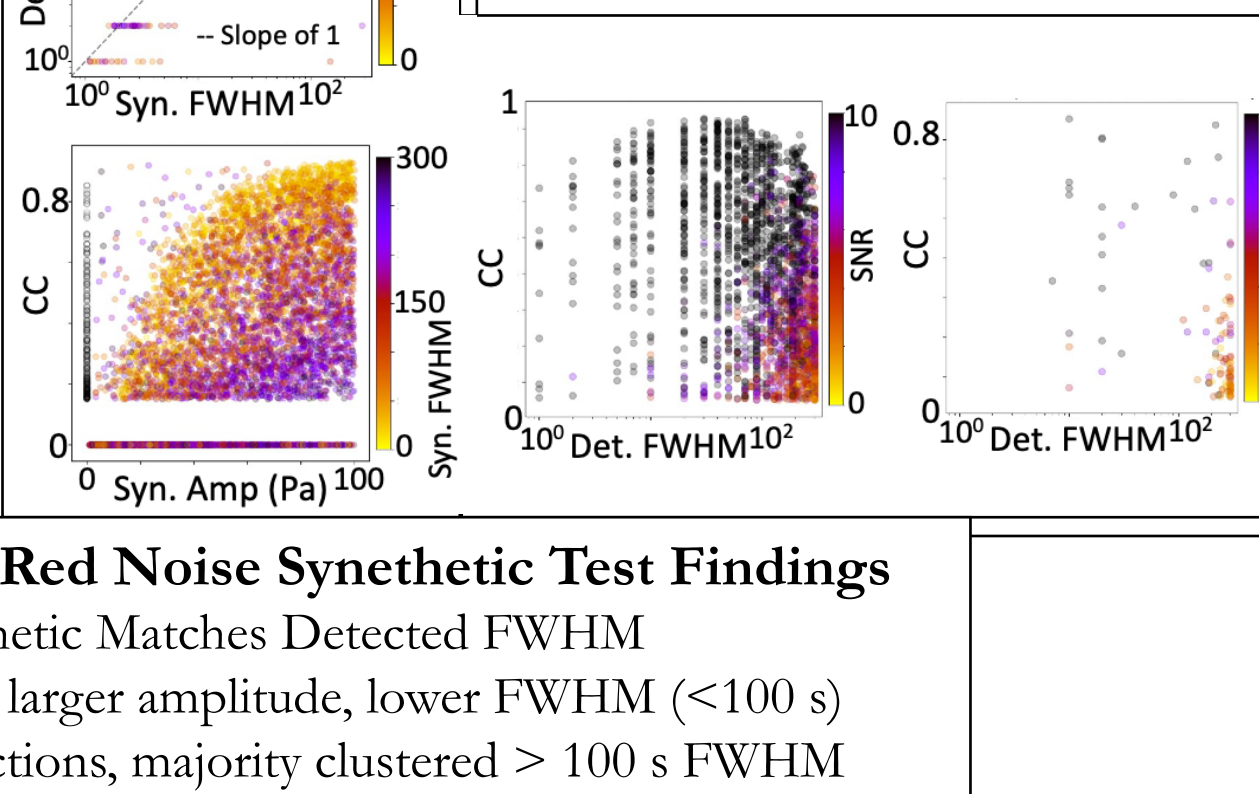
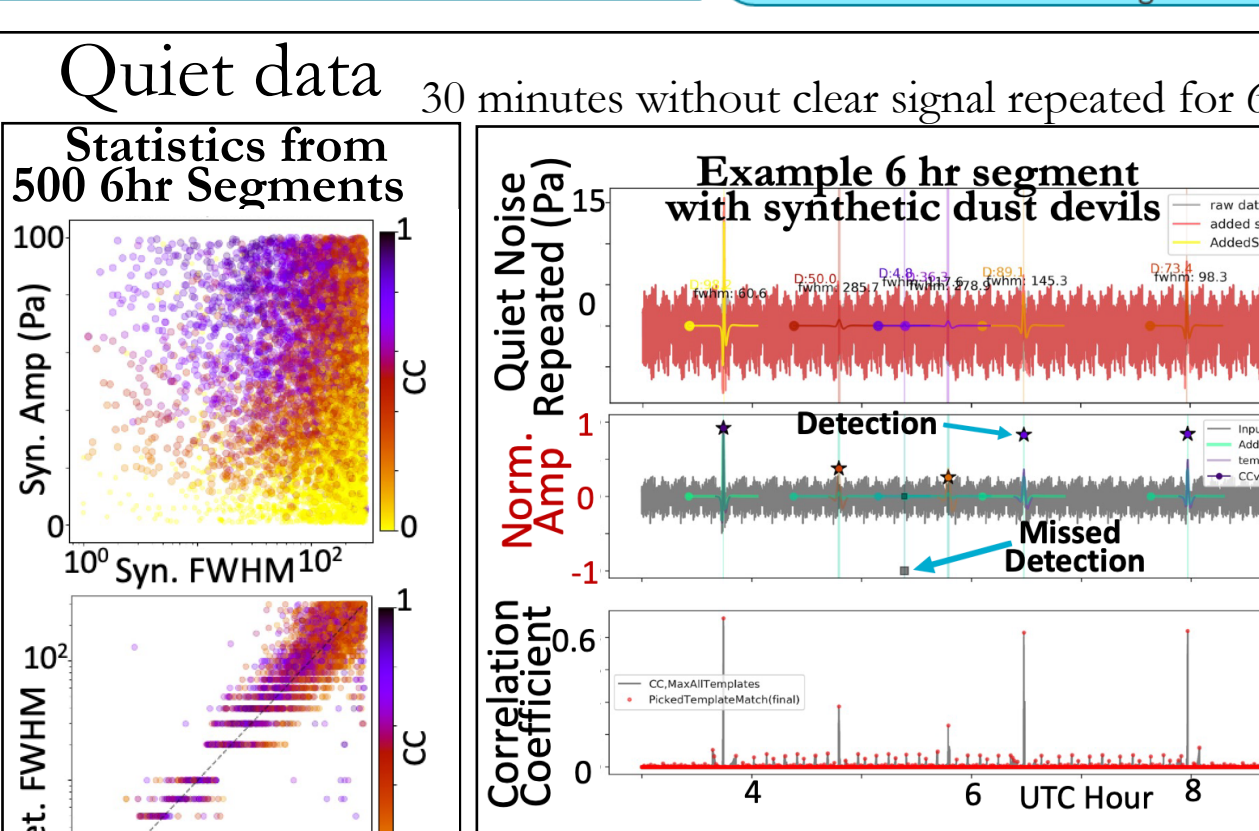


Red Noise Synthetic Data for D5M04

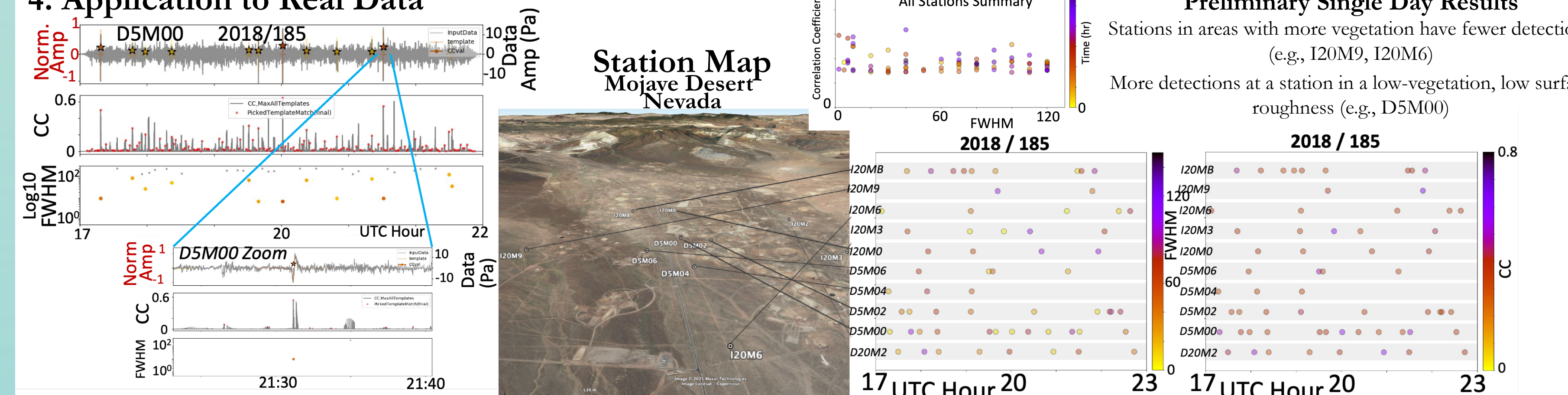


3. Synthetic Tests

Set up Empty 'Data'
Quiet data (500 6 hr Segments)
Red Noise (250 6 hr Segments)



4. Application to Real Data



Wavelet Detector

Goal: Distinguish peaks due to dust devils from background noise via wavelets

Methodology

χ^2 statistical significance testing of wavelet power at time index n and scaling s , $|W_n(s)|^2$ to detect dust devils:

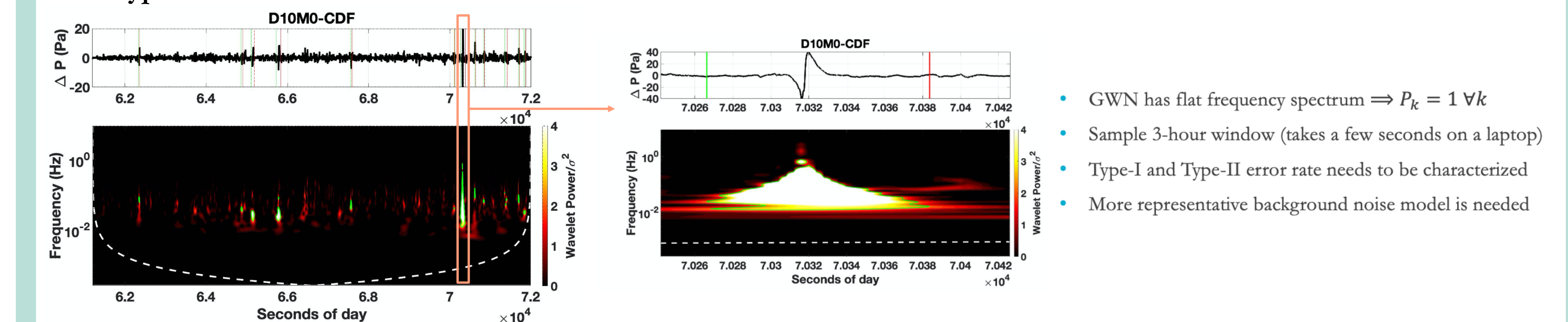
$$\frac{|W_n(s)|^2}{\sigma^2 P_k} \rightarrow \frac{\chi_k^2}{2} \quad \text{Torrence and Compo (1997)}$$

where P_k is noise power in the k^{th} frequency bin and σ^2 is the signal variance

Processing

1. Decimate time series from 1000 Hz to 25 Hz sampling rate through a 2-stage decimation process. Final filter cutoff is 10 Hz.
2. Window the signal into chunks (~ few hours)
3. Generate wavelet coefficients using continuous wavelet transform with Morlet wavelets
4. Compute outlier-removed variance of the signal and noise power spectrum for normalization of wavelet power
5. Compare normalized wavelet power with values of $\chi_k^2(1 - \alpha)$, with α as the required confidence interval.
6. Mark start and end times of regions that have normalized wavelet power greater than $\frac{1}{2}\chi_k^2(1 - \alpha)$

Prototype with White Gaussian Noise



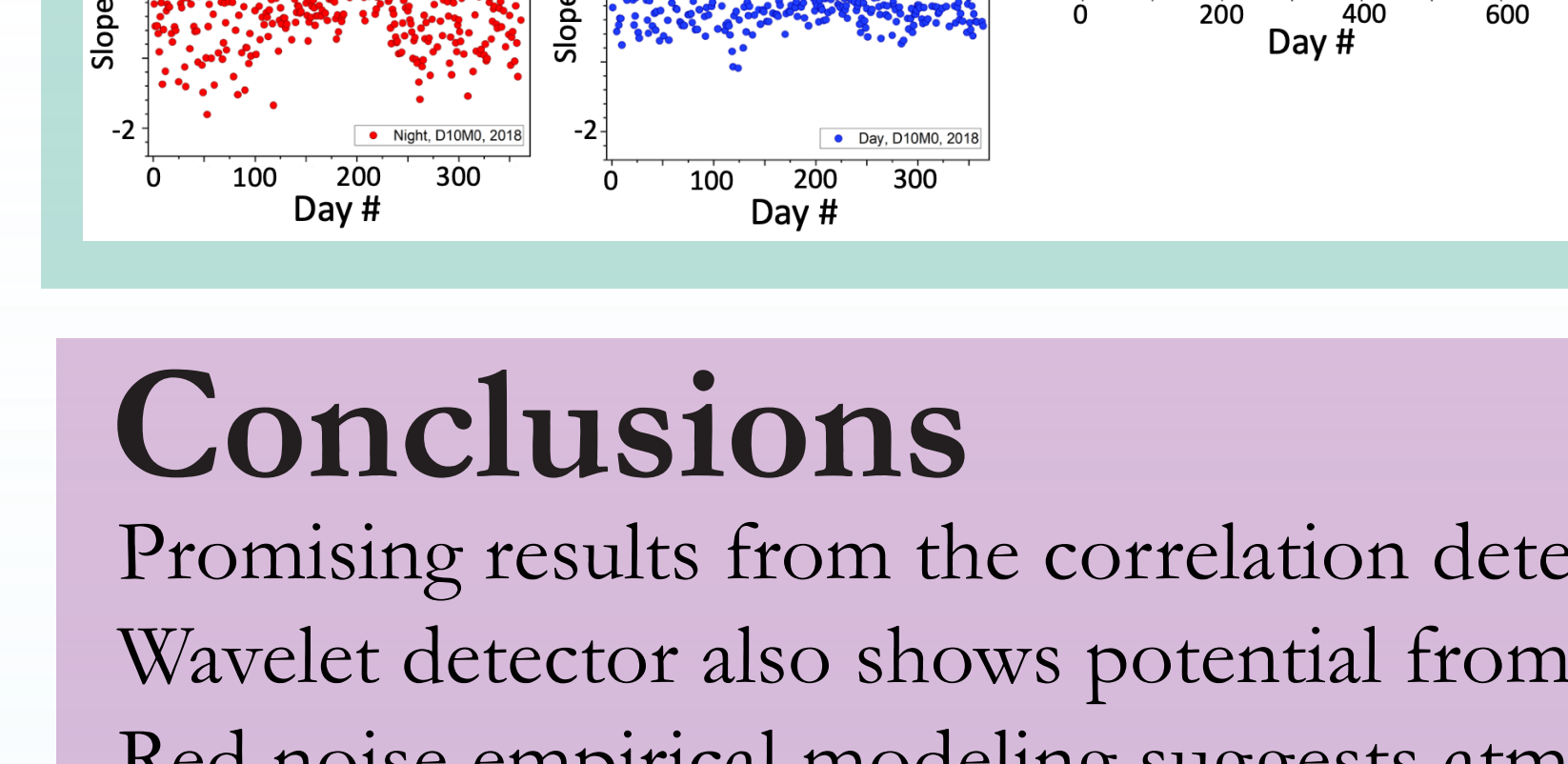
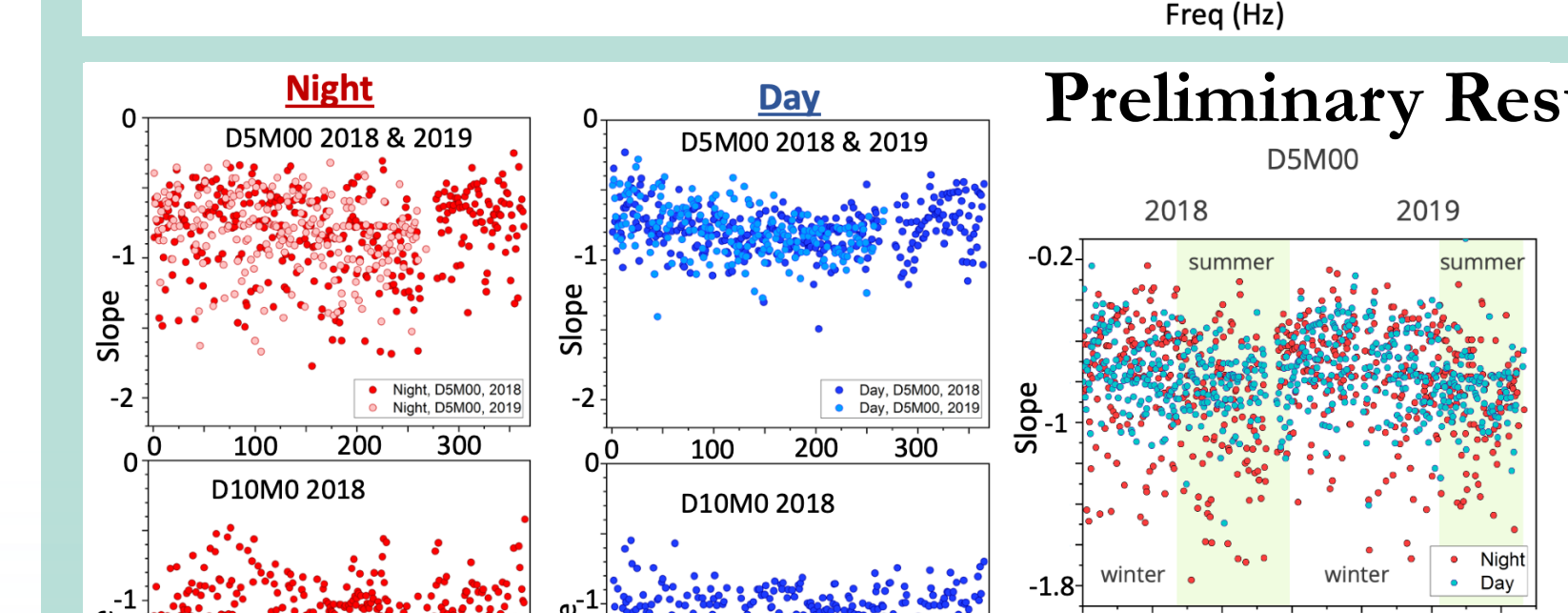
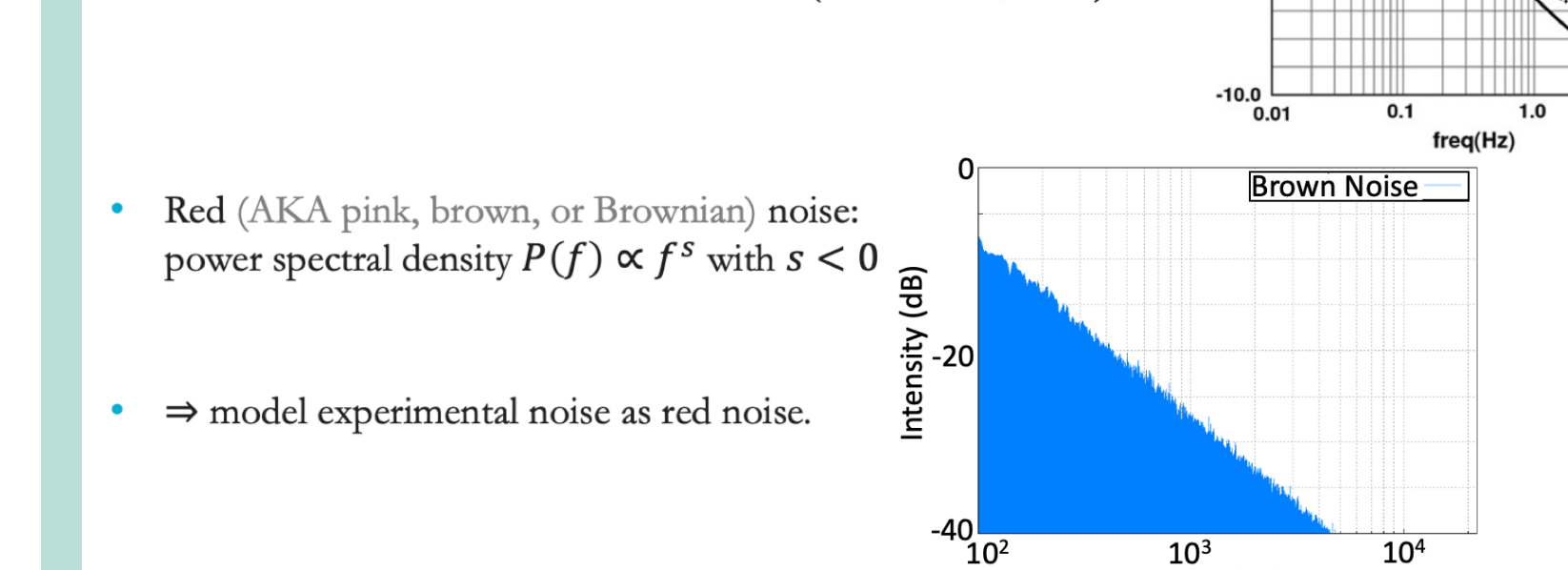
- GWN has flat frequency spectrum $\Rightarrow P_k \approx 1 \forall k$
- Sample 3-hour window (takes a few seconds on a laptop)
- Type-I and Type-II error rate needs to be characterized
- More representative background noise model is needed

Red Noise Empirical Modeling

Goal: Determine background noise model from stations spanning multiple years

Methodology

- Atmospheric dynamics follows energy cascade: larger eddies dissipate to smaller eddies.
- Pressure fluctuations (e.g., infrasound) follow same pattern. See IDC Global Infrasound Noise Model (Brown *et al.*, 2014).



Key Findings

Larger variation of slope values during night hours in comparison to day

Lower absolute slope values a winter, although strong distribution, than summer

Constant at night lower values than during the day

Constant also sees range of values that will be investigated moving forward

Conclusions

Promising results from the correlation detector, will be extended to more stations across greater number of days

Wavelet detector also shows potential from initial white Gaussian noise application

Red noise empirical modeling suggests atmospheric complexities that will be investigated

References

- Brown, D., Ceranna, L., Prior, M., Mialle, P. & Le Bras, R. J. (2014) Pure Appl. Geophys. 171, 361–375. doi: 10.1007/s00024-012-0573-6
- Gibbons, S. J., & Ringdal, F. (2006). Geophys. J. Int., 165(1), 149–166. doi: 10.1111/j.1365-246X.2006.02865.x
- Jackson, B., Lorenz, R. D. and Davis, K. (2018). Icarus 299 166–174, doi: 10.1016/j.icarus.2017.07.027
- Jackson, B. and Lorenz, R. D. (2015). J. Geophys. Res. Planets 120, doi: 10.1002/2014JE004712
- Lorenz, R. D. and Jackson, B. K. (2015). GeoResJ 1–11, doi: 10.1785/1020150133/
- Lorenz, R. D. and Jackson, B. K. (2016). Space Sci. Rev. 203 277–297, doi: 10.1007/s11214-016-0277-9
- Lorenz, R. D., Balme, M. R., Gu, Z., Kahanpää, H., Klose, M., Kurgansky, ... and Wei, W. (2016). Space Sci. Rev. 203 5–37, doi: 10.1007/s11214-016-0239-2
- Torrence, C. and Compo, G. P. (1998). Bull Am Meteorol Soc 79 (1) 61–78, doi: 10.1175/15200477(1998)079<0061:APGTWA>2.0.CO;2