

Performance Evaluation of a Shaped Sonic Boom Detector and Classifier

Blaine M. Harker, Shane V. Lympny, Juliet A. Page
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Community Noise Testing



>100
Recorders



Reliable
Detections



Rapid
Analysis



Outline

1. Review Community Noise Testing Workflow
2. Shaped Sonic Boom Detector
3. Shaped Sonic Boom Classifier
4. Preliminary Performance Results
5. Summary



Outline

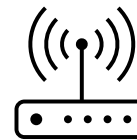
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Community Noise Measurements



Begin Recording

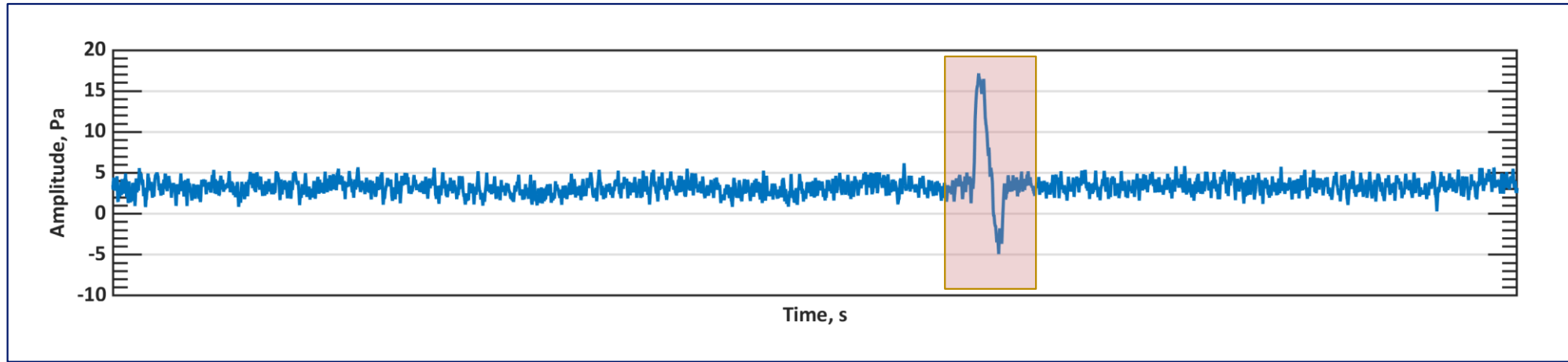


Host Station



Recording
Station

Community Noise Measurements



Recording Station

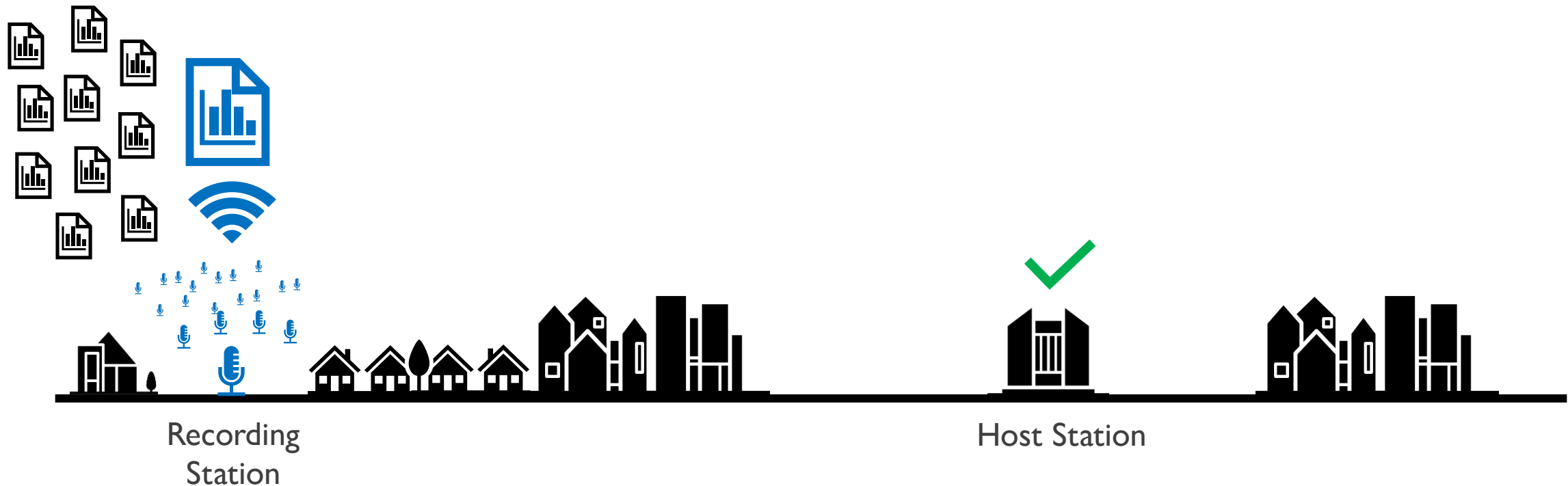
Host Station

Community Noise Measurements



▶ Recording Station Processing Software

- Sonic Boom Detector
- Sonic Boom Classifier
- Sonic Boom Metrics Calculations



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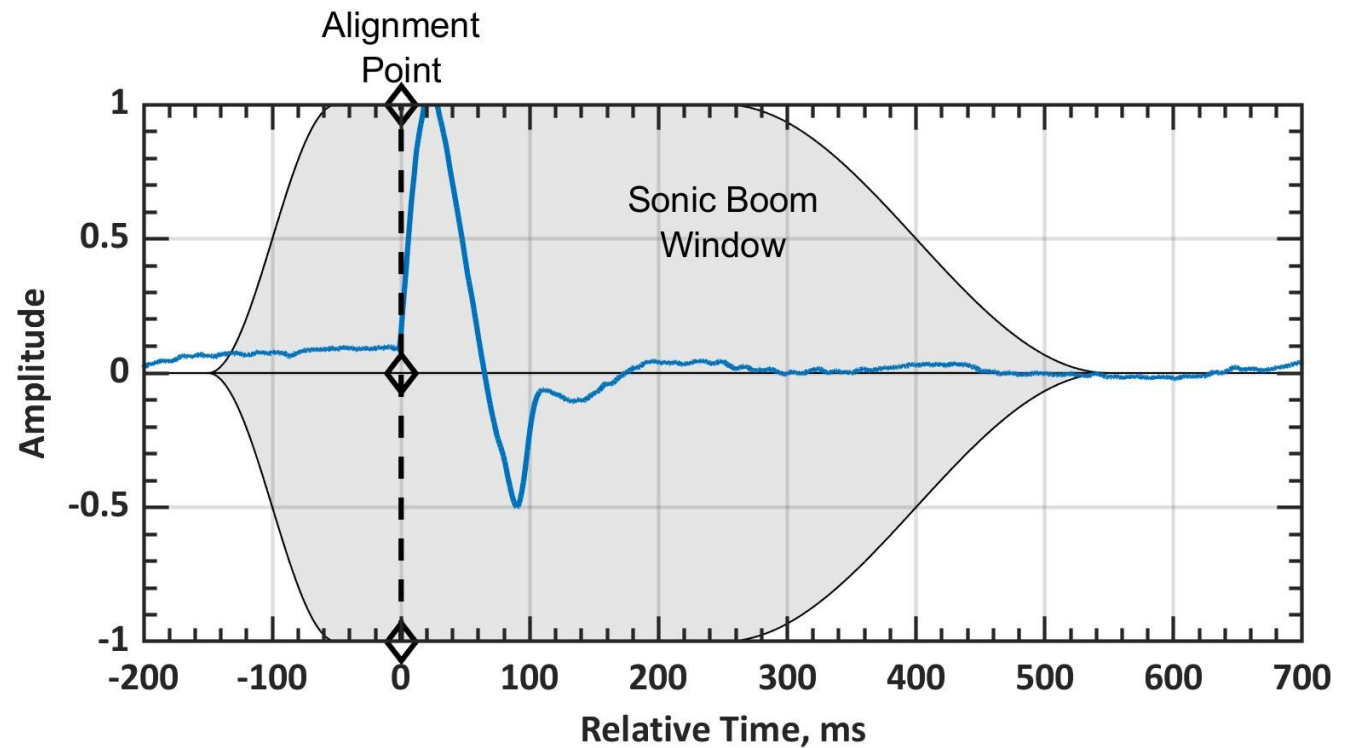
Shaped Sonic Boom Detector



- ▶ Goal: Detect sonic boom with temporal accuracy of <100 ms



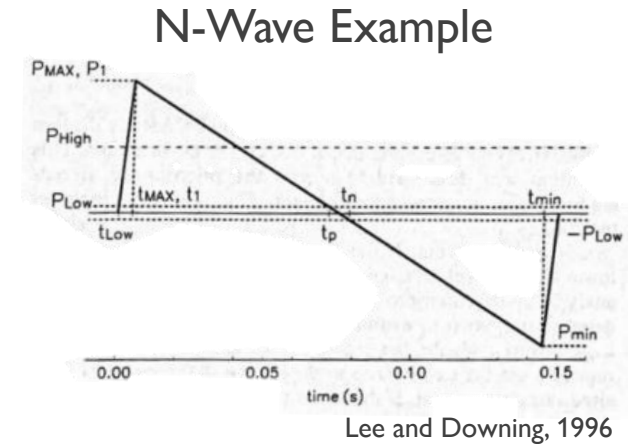
- ▶ Local detection time used to window and calculate sonic boom metrics



Select Prior Work



- ▶ BEARS Algorithm (Lee and Downing, 1996)
 - Rise-time-based method for N-waves
- ▶ Auto Boom Finder (Hobbs, 2012)
 - Bandpass amplitude detector for N-waves
 - Utilized in WSPR 2011 and QSF18
- ▶ Spectral Fingerprint Method (Klos, 2022)
 - Compares the spectrogram of a simulated shaped sonic boom to the waveform spectrogram
 - Highly effective for nontraditional booms



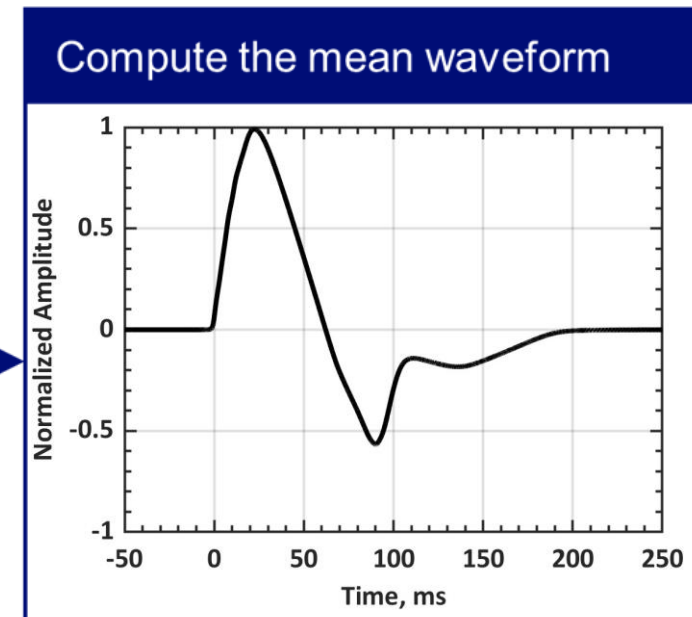
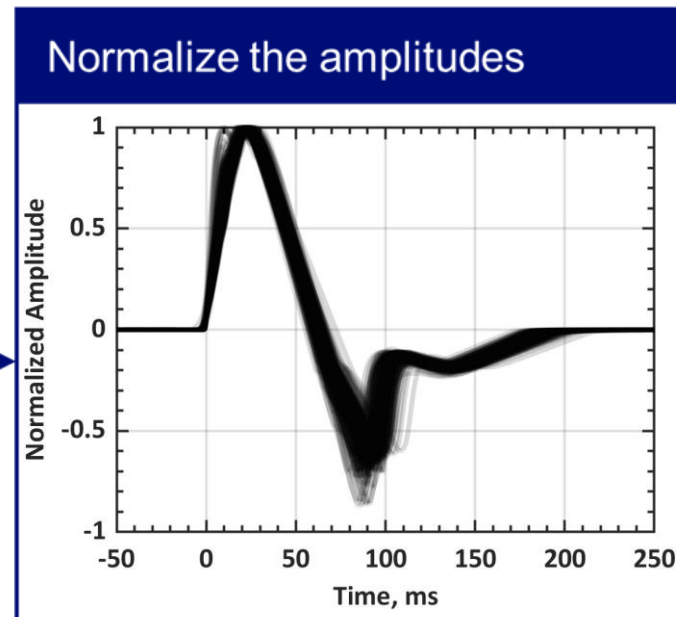
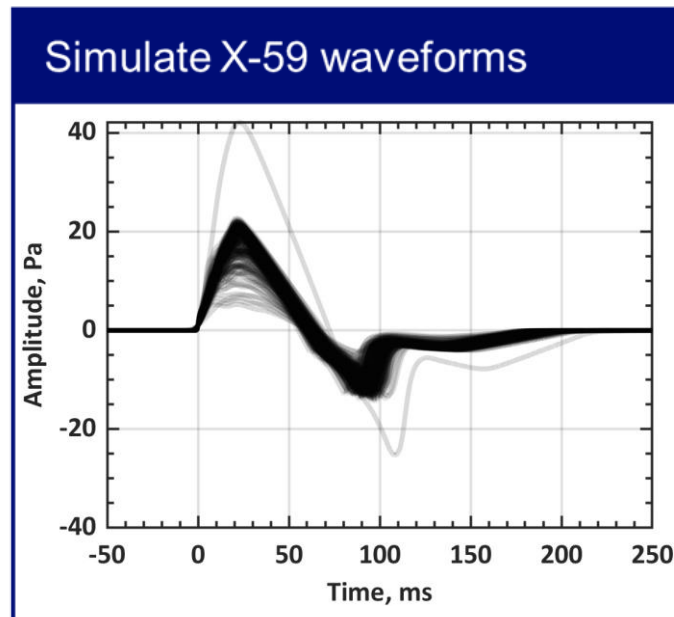
Lee and Downing, "Boom Event Analyzer Recorder: Unmanned Sonic Boom Monitor", J Aircraft 33 (1), pp. 171-175 (1996).

Hobbs, "Auto Boom Finder Program (ABF)," Wyle Technical Note TN 12-30, Arlington, Va (2012).

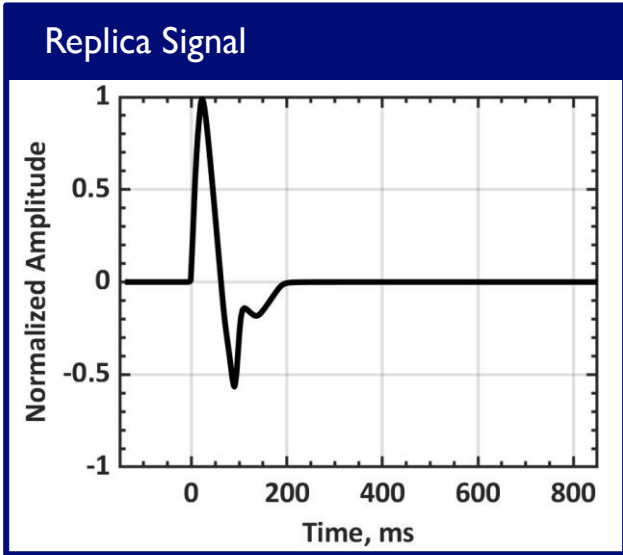
Klos, "Finding X-59: A Spectral Fingerprint Based Sonic Boom Finder Algorithm" NASA TM (Unpublished Draft) (2022).

Replica Signal Generation

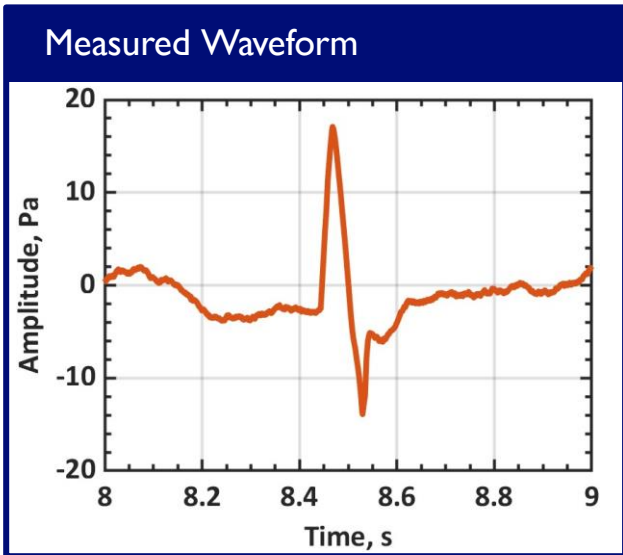
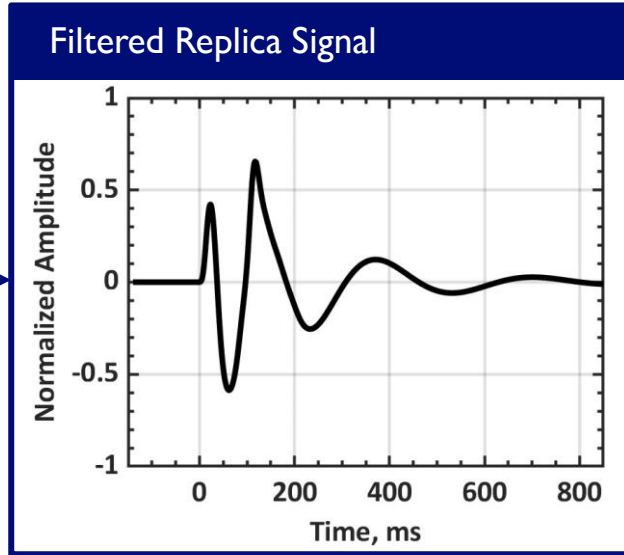
- ▶ Replica signal is chosen to most resemble the expected sonic boom waveform
- ▶ 200 X-59 Sonic Boom Waveforms (W. Doebler) from C612A (On Design) and propagated using PCBoom



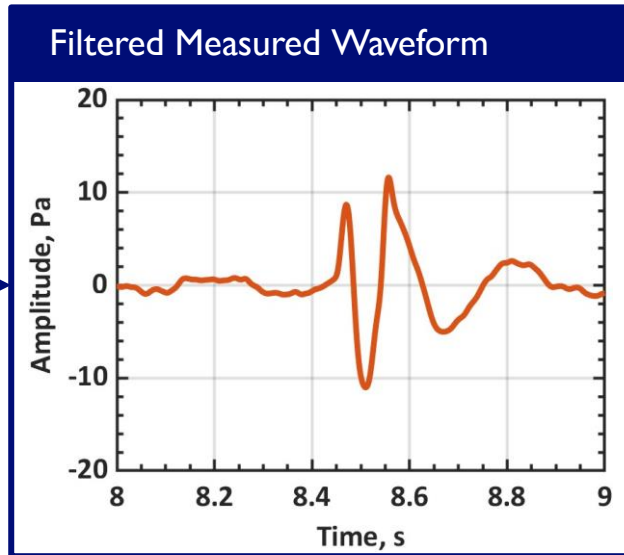
Replica Correlator



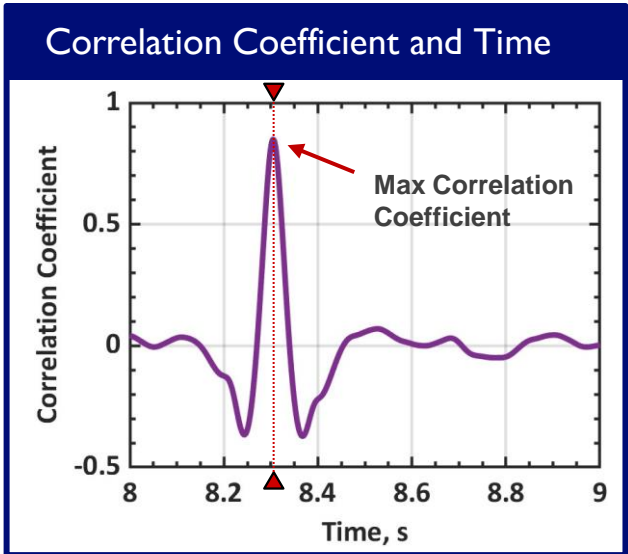
Filter



Filter



Cross Correlation



Outputs

Max Correlation Coefficient

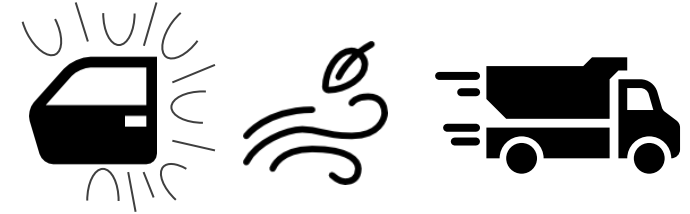
Detection Timestamp

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Sonic Boom Classifier



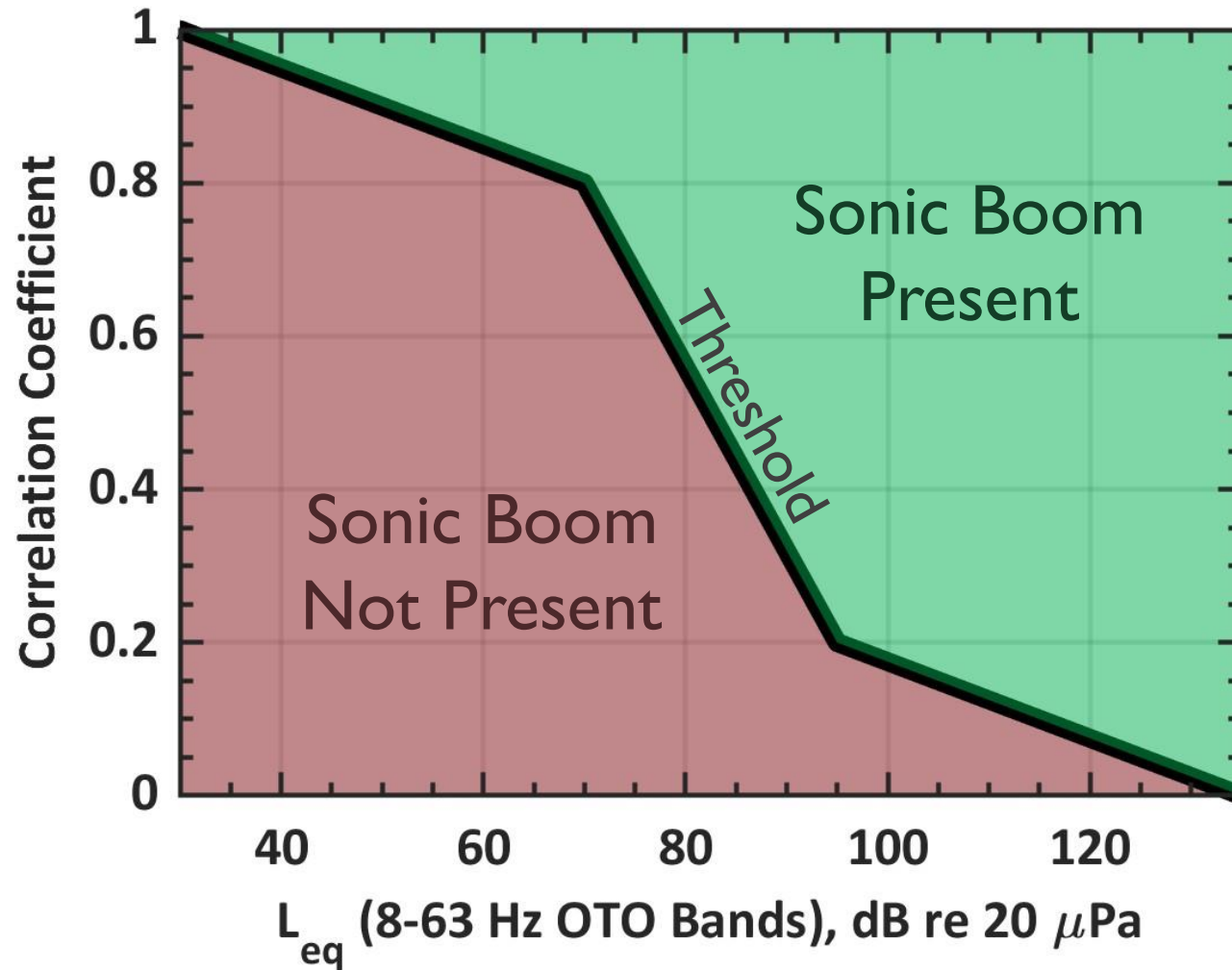
Why do we need a classifier?

- ▶ No guarantee that a GRS will record a sonic boom, but the detector will output the most likely sonic boom detection time
- ▶ Need a method to classify a sonic boom in the presence of ambient noise

Key Parameters

- ▶ Detector correlation coefficient
- ▶ Corresponding sonic boom level (select OTO band levels)

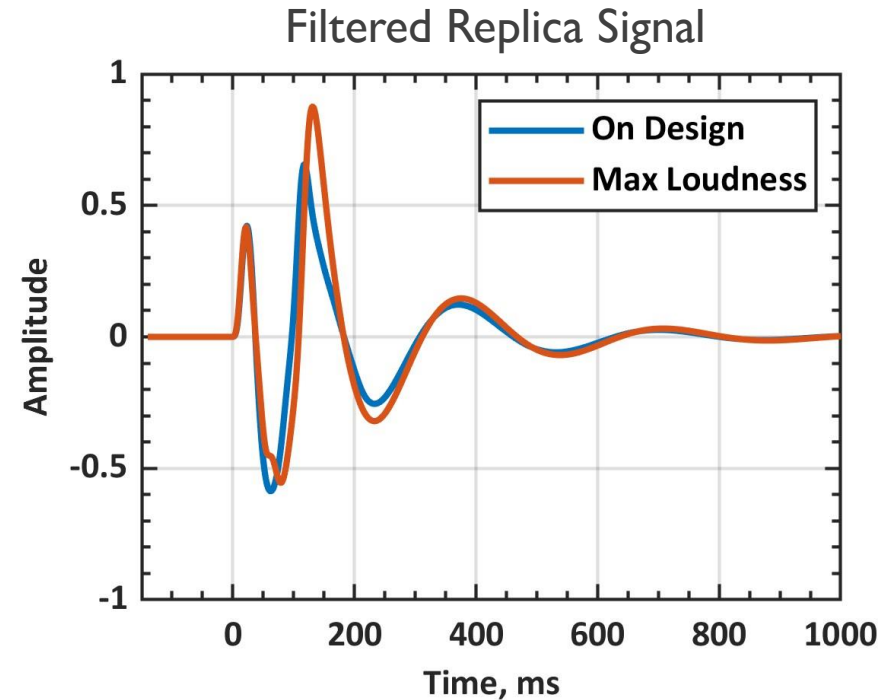
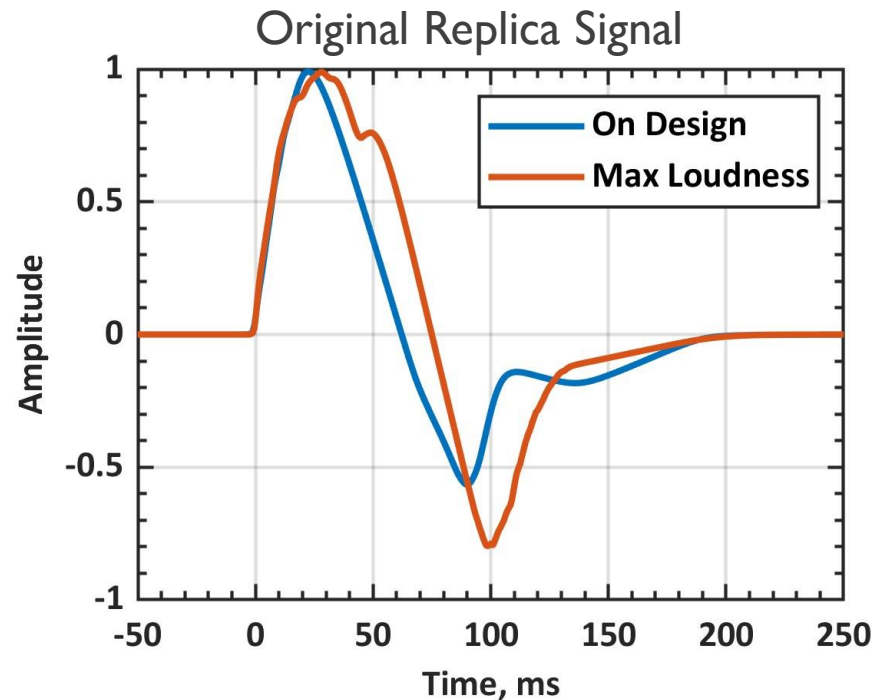
Classifier Description



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- **Sonic Boom Simulations**

- NASA C612A Cylinder comprises two designs: On Design (Min Loudness) and Max Loudness (Off Design)
- Propagated in various undertrack scenarios via PCBoom for ~10k samples each
- Replica signal will use On-Design case

NASA-Provided Datasets



- **Dataset:**

- 25,000 samples per set
- 30-second waveforms
- X-59 Sonic boom propagated via PCBoom
- Ambient noise from previous measurement (Galveston, TX)

Dataset	Turbulence	Post boom noise	Impulsive noises
Set 1	No	No	No
Set 2	Yes	No	No
Set 3	Yes	Yes (+0 dB)	No
Set 4	Yes	Yes (–10 dB)	No
Set 5	Yes	Yes (–10 dB)	Yes

Ref: Table 3 of Klos (2022)

Sonic Boom Detector Performance – On Design

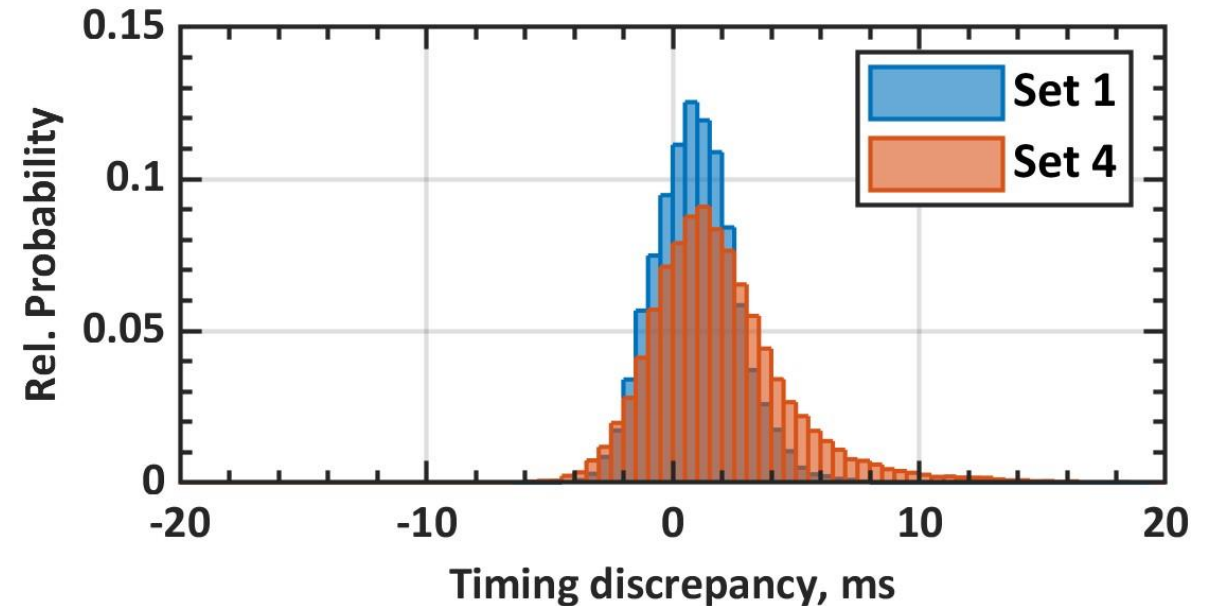


- ▶ Detection errors in each dataset were relatively similar
- ▶ Detection Failure:
|Timing discrepancy| > 100 ms
- ▶ Failure rate was ~0.01% (1:10000)
- ▶ Detection times were within ± 20 ms of actual time in 99.8% of tests

Failure Rates	Set 1	Set 2	Set 3	Set 4	Set 5
Detection > ± 20 ms	0.01%	0.16%	0.18%	0.18%	0.16%
Detection > ± 100 ms	0.01%	0%	0.01%	0.01%	0.01%

Sonic Boom in Ambient Noise

Sonic Boom in Ambient Noise + Post-boom noise



Sonic Boom Detector Performance – Max Loudness

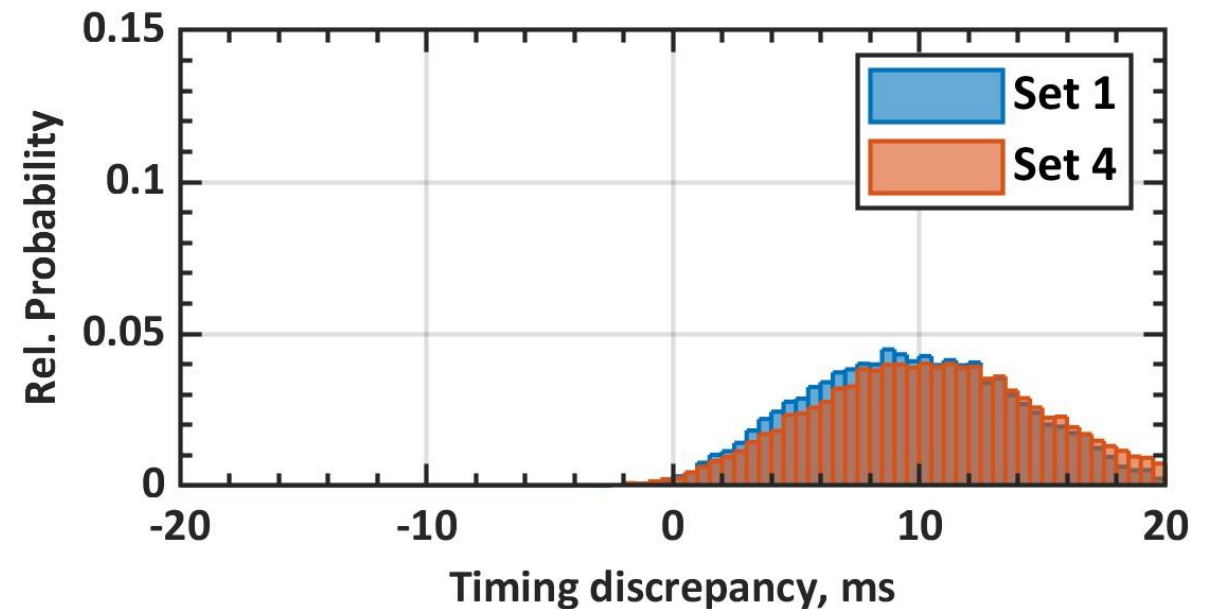


- ▶ Detection errors in each dataset were relatively similar
- ▶ Detection Failure:
|Timing discrepancy| > 100 ms
- ▶ Failure rate was ~0.01% (1:10000)
- ▶ Detection times were within ± 20 ms of actual time in 95-99% of tests

Failure Rates	Set 1	Set 2	Set 3	Set 4	Set 5
Detection > ± 20 ms	0.48%	4.17%	4.25%	4.16%	4.18%
Detection > ± 100 ms	0%	0.01%	0.01%	0%	0.01%

Sonic Boom in Ambient Noise

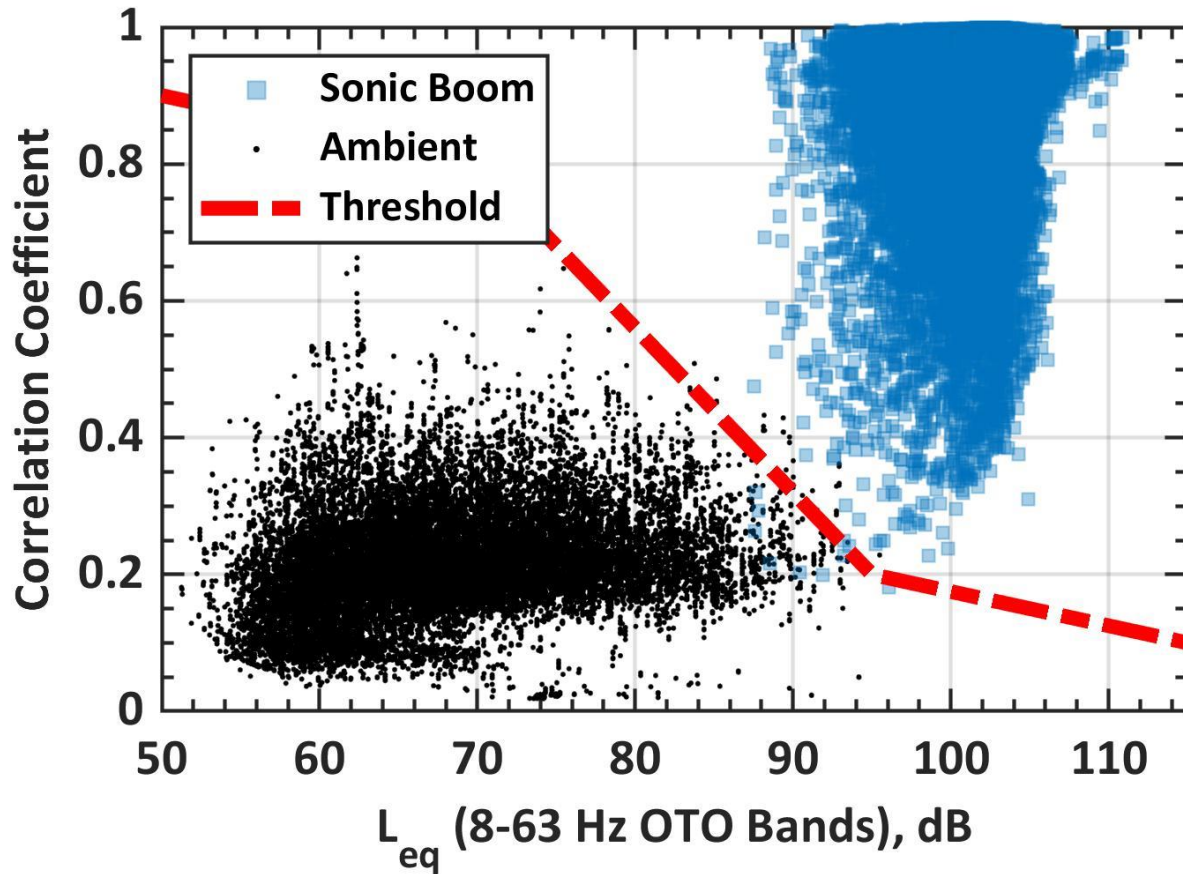
Sonic Boom in Ambient Noise + Post-boom noise



Classifier Performance

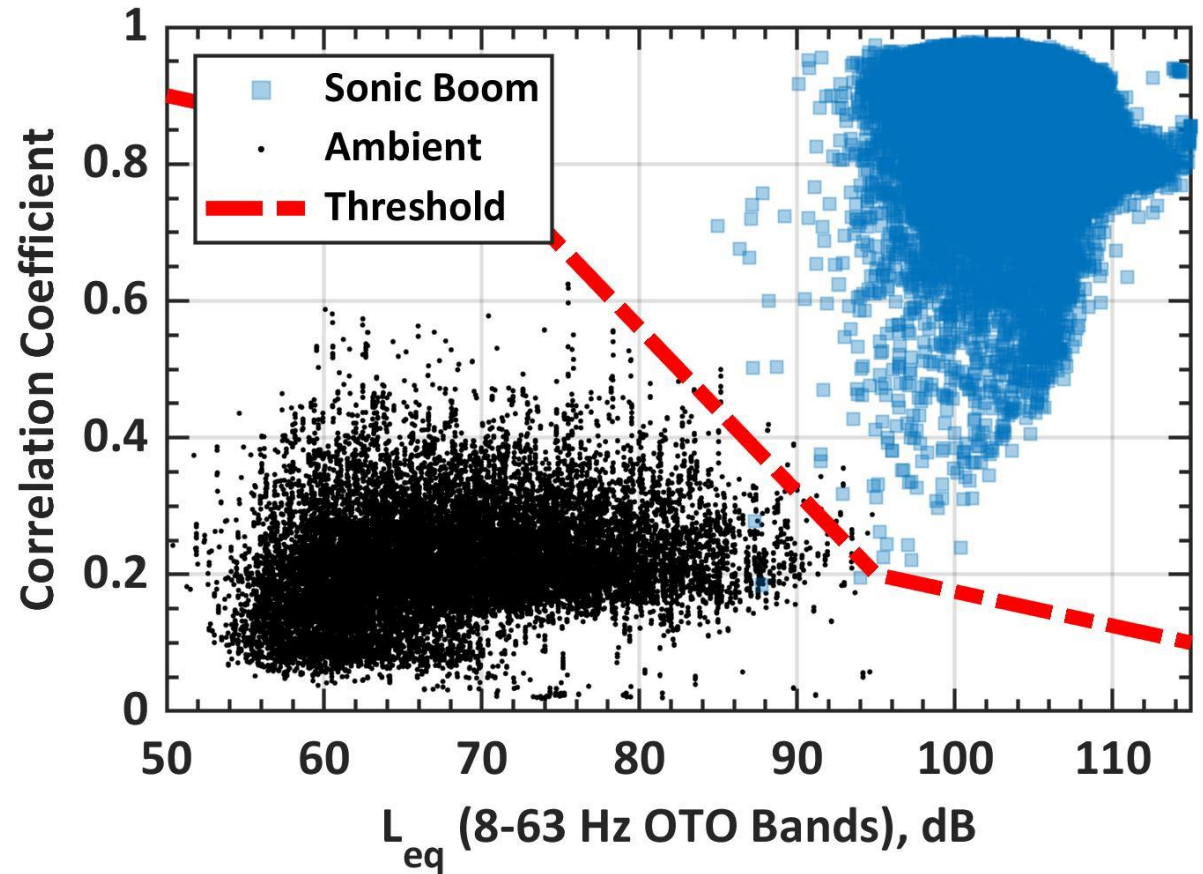


Classifier - On Design Cases



Classification Performance Results
 True Positive Rate: .9999
 False Positive Rate: .0024

Classifier Max Loudness Cases



Classification Performance Results
 True Positive Rate: >.9999
 False Positive Rate: .0023

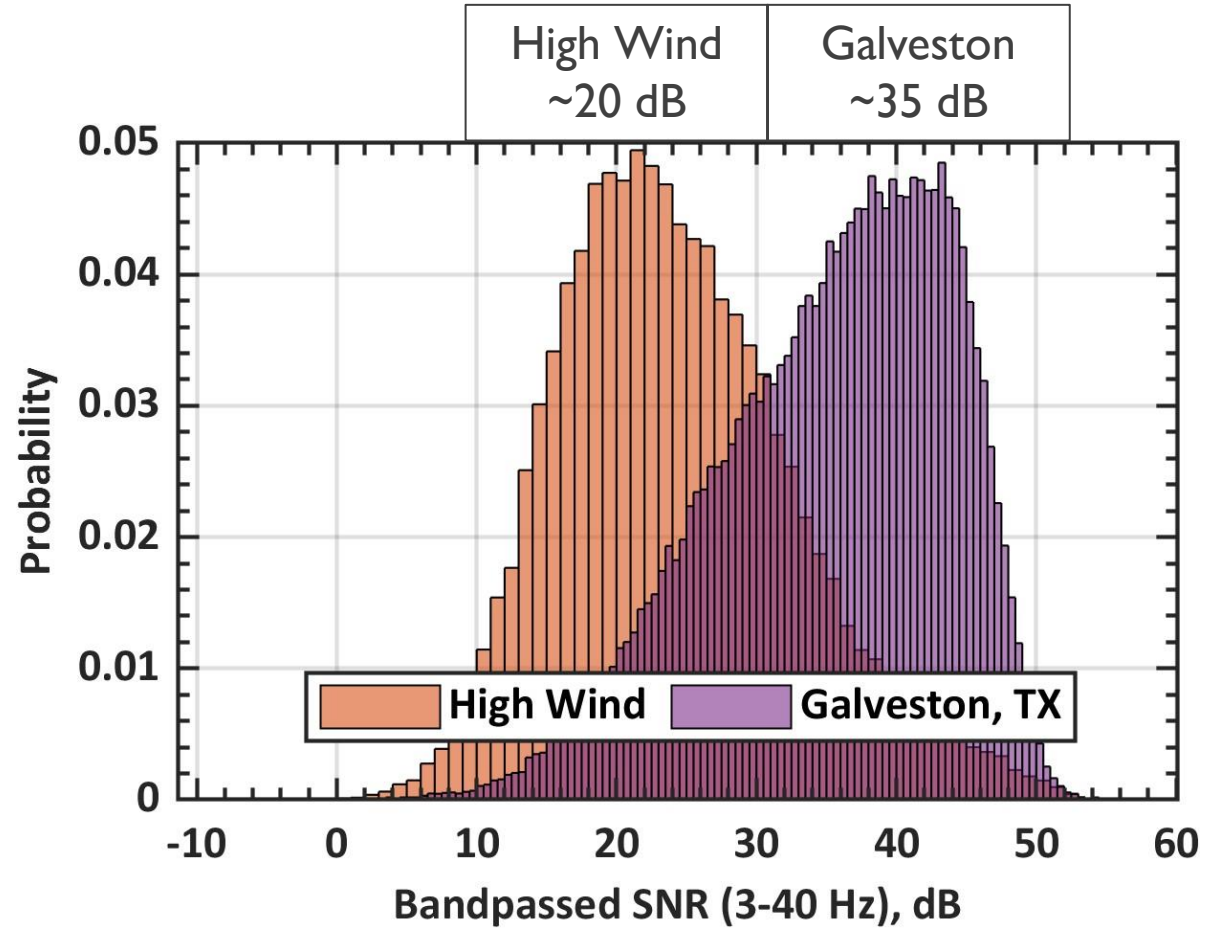
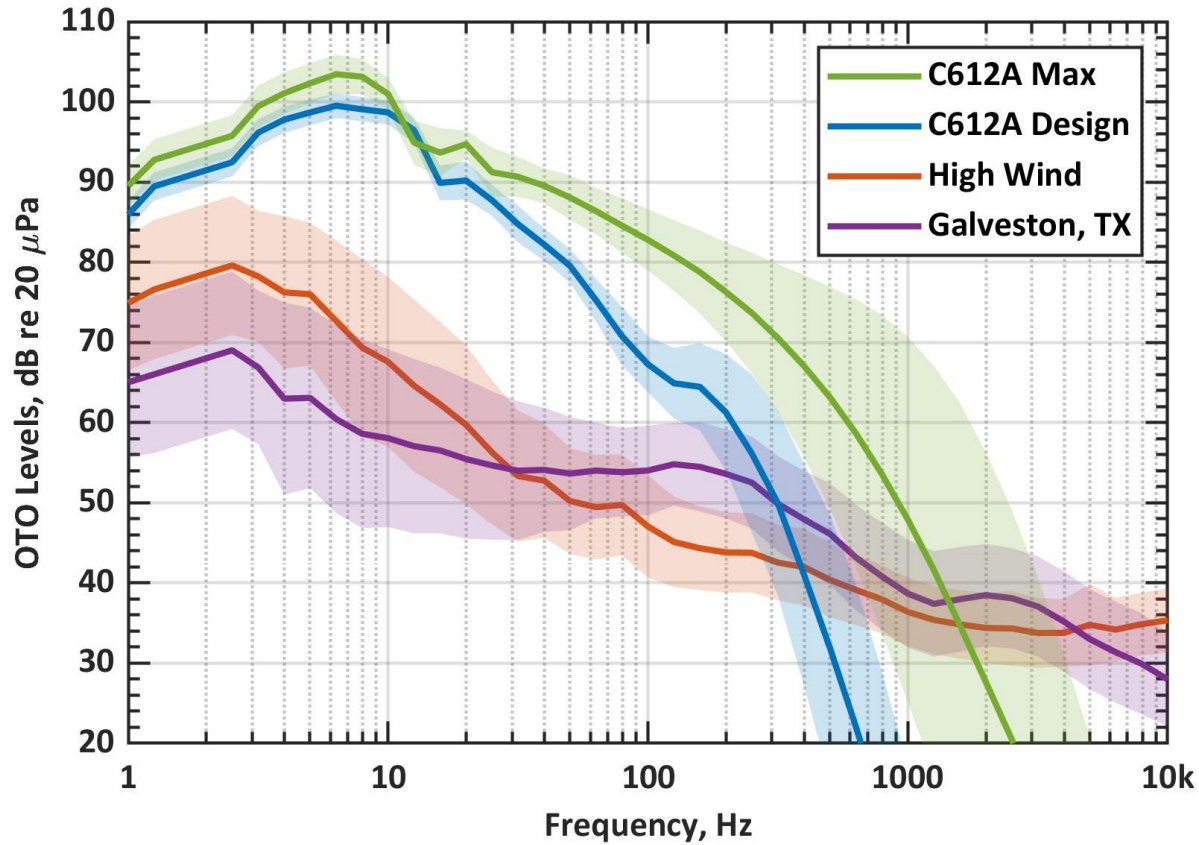
Detector performance in high-wind environments



- ▶ Detector relies on high SNR in low-frequency bands
- ▶ High-wind ambient recordings
 - 15-22 mph sustained winds
 - 27-38 mph gusts
- ▶ ~20 min of ambient recordings
 - Ground microphones
 - Large windscreens (23 cm diameter)



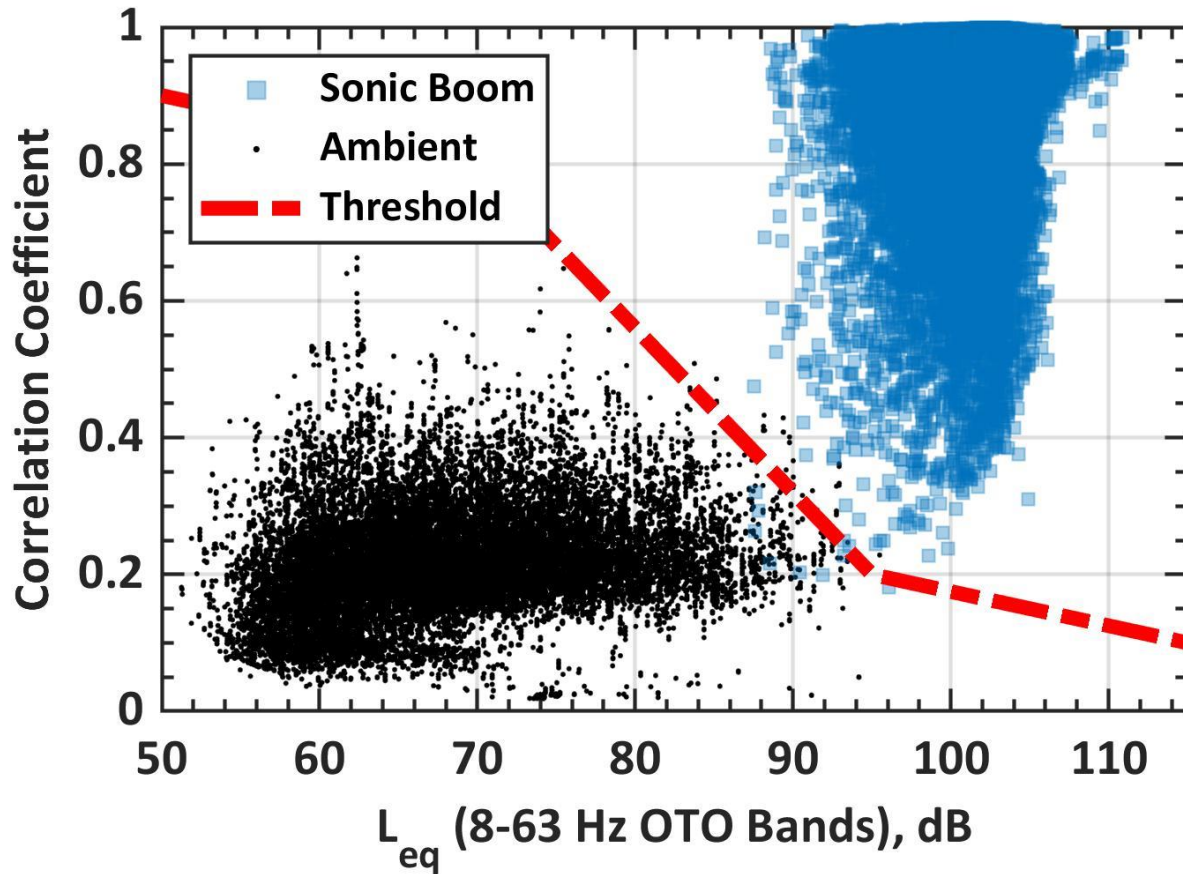
X-59 Sonic Boom and Ambient Noise Spectra



Signal to Noise Ratio for Different Noise Datasets

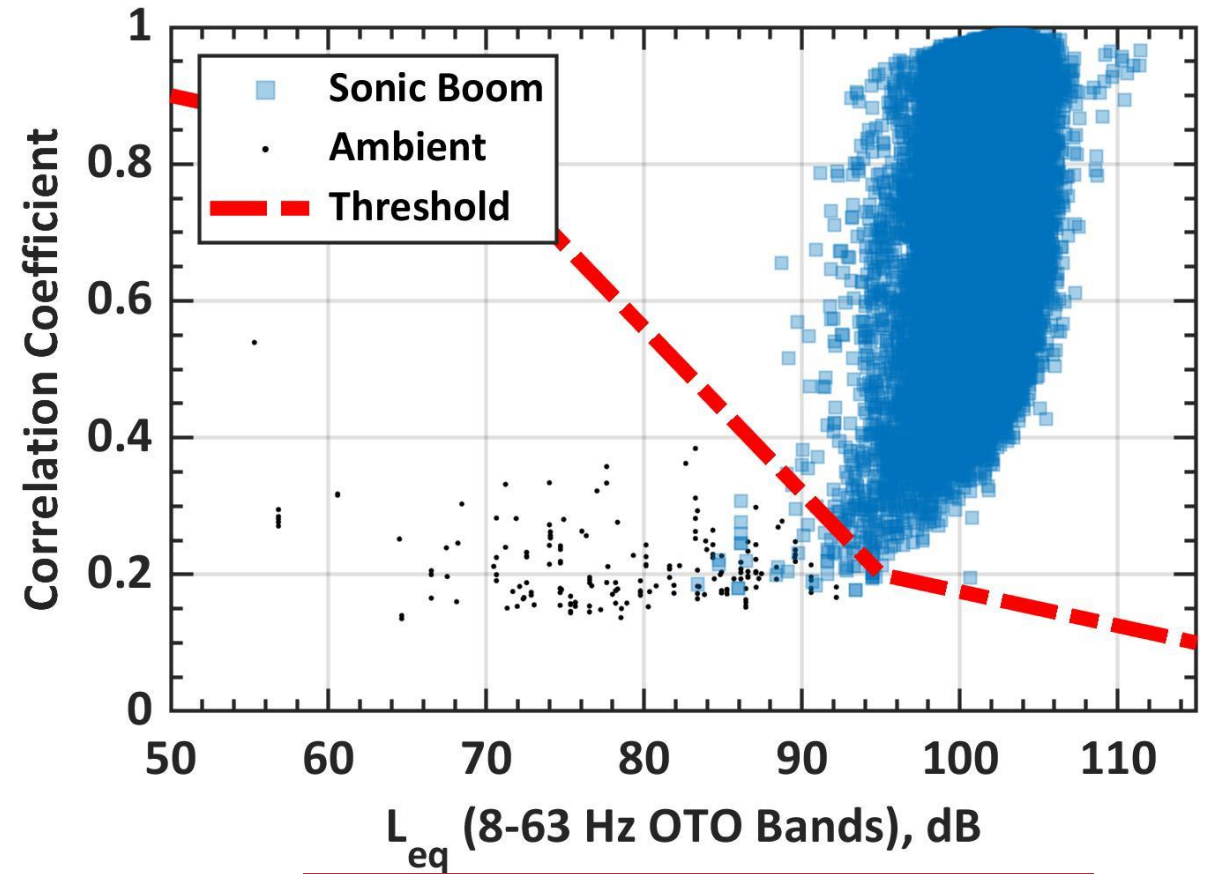


Classifier – On Design – Galveston



Classification Performance Results
 True Positive Rate: .9999
 False Positive Rate: .0024

Classifier – On Design – High Wind



Classification Performance Results
 True Positive Rate: .9989
 False Positive Rate: 0 (Data Limited)

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Shaped Sonic Boom Detector

- ▶ Replica correlator design
- ▶ Replica signal based on simulation—generalizable to any signal
- ▶ Failure rate 1:10,000

Shaped Sonic Boom Classifier

- ▶ Input detection correlation coefficient *and* bandlimited levels at detection
- ▶ True positive rate ~ 0.9999 with False Positive Rate ~ 0.0024
- ▶ Expect >20 dB SNR in frequency band of interest, 35 dB SNR Typical

Backup Slides



ROC Curves for Different Ambient Noise Datasets

