

# **Spectrum Forensics: New Observations and Analytics**

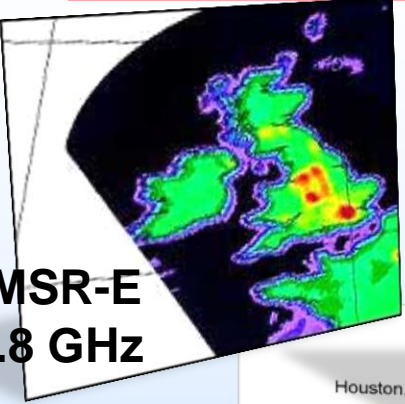
Prof. Albin J. Gasiewski

Center for Environmental Technology  
Department of ECEE  
University of Colorado  
Boulder, CO

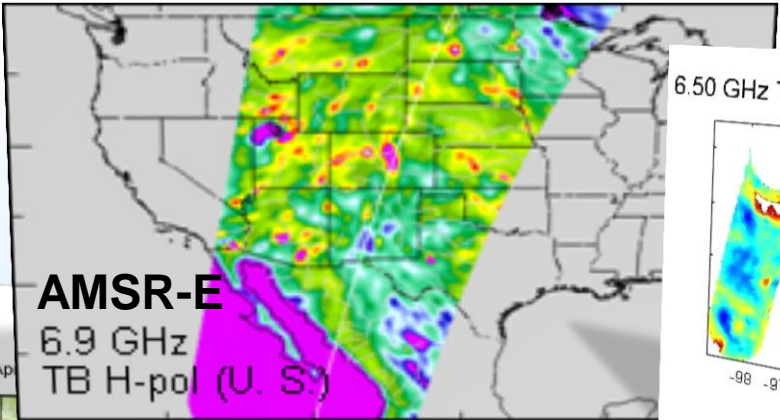
[al.gasiewski@colorado.edu](mailto:al.gasiewski@colorado.edu)

# Anthropogenic RFI in Passive Microwave Measurements

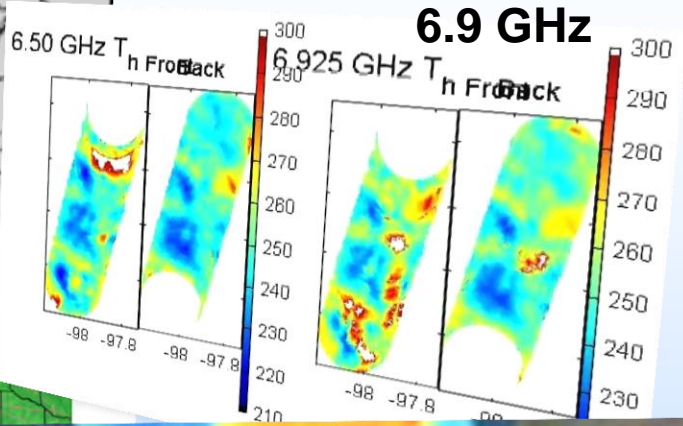
**AMSR-E  
6.8 GHz**



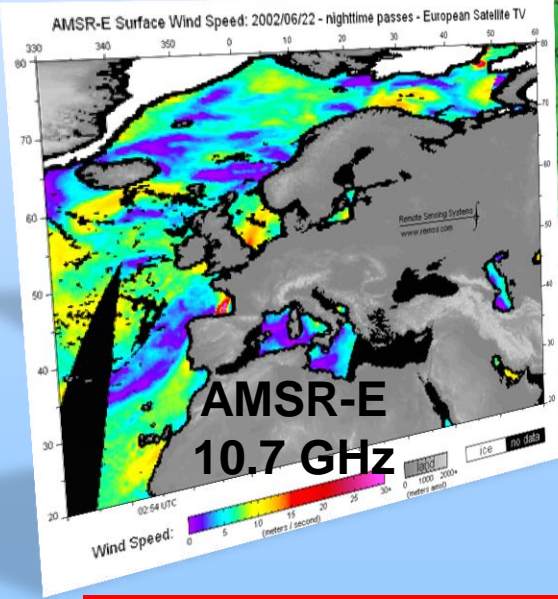
**AMSR-E  
6.9 GHz  
TB H-pol (U. S.)**



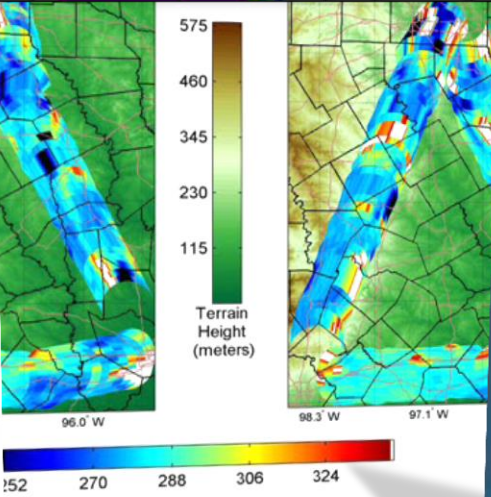
**PSR/C  
6.9 GHz**



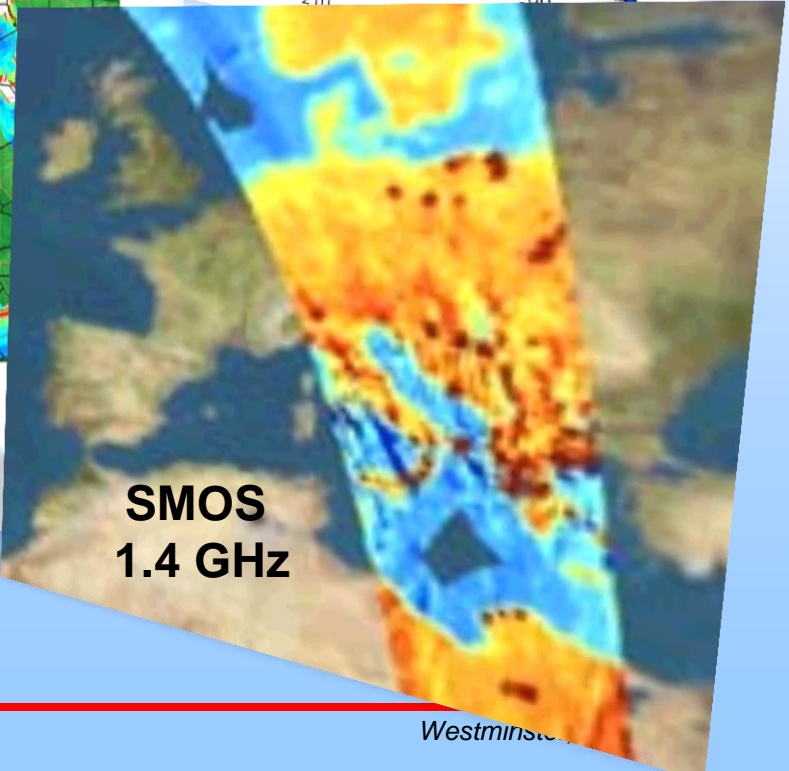
**AMSR-E  
10.7 GHz**



**PSR/CXI  
6.9 GHz**



**SMOS  
1.4 GHz**

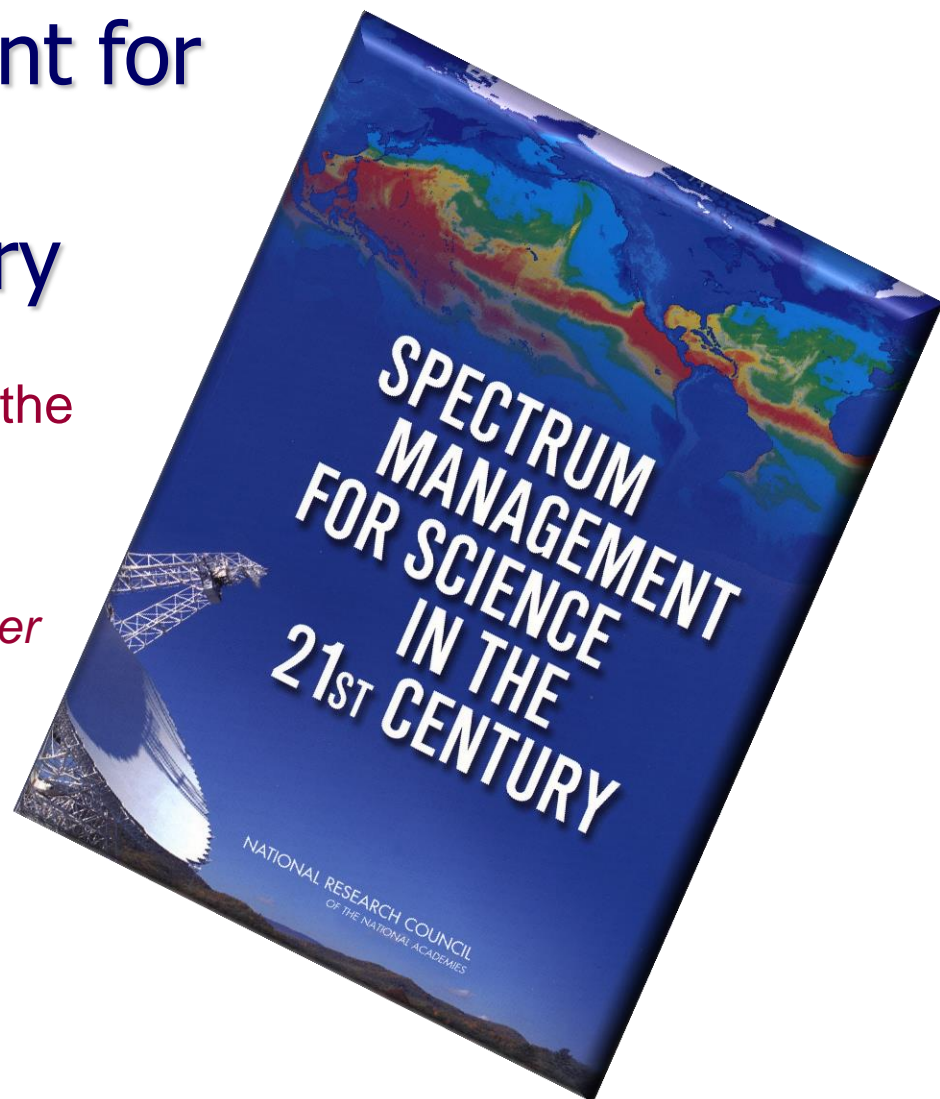


# Spectrum Management for Science in the 21st Century

Committee on Scientific Uses of the  
Radio Spectrum

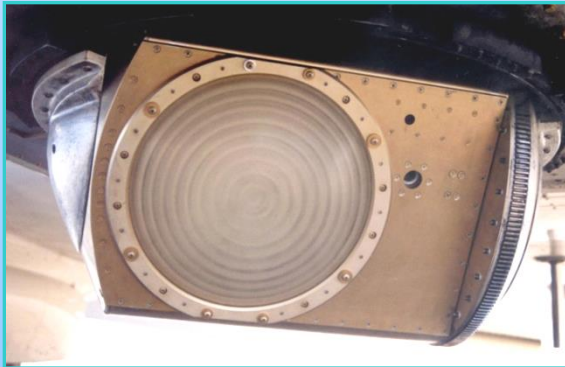
David B. Lang *NRC Program Officer*  
Albin J. Gasiewski, *Co-Chair*  
Marshall H. Cohen, *Co-Chair*

*on behalf of the full committee*





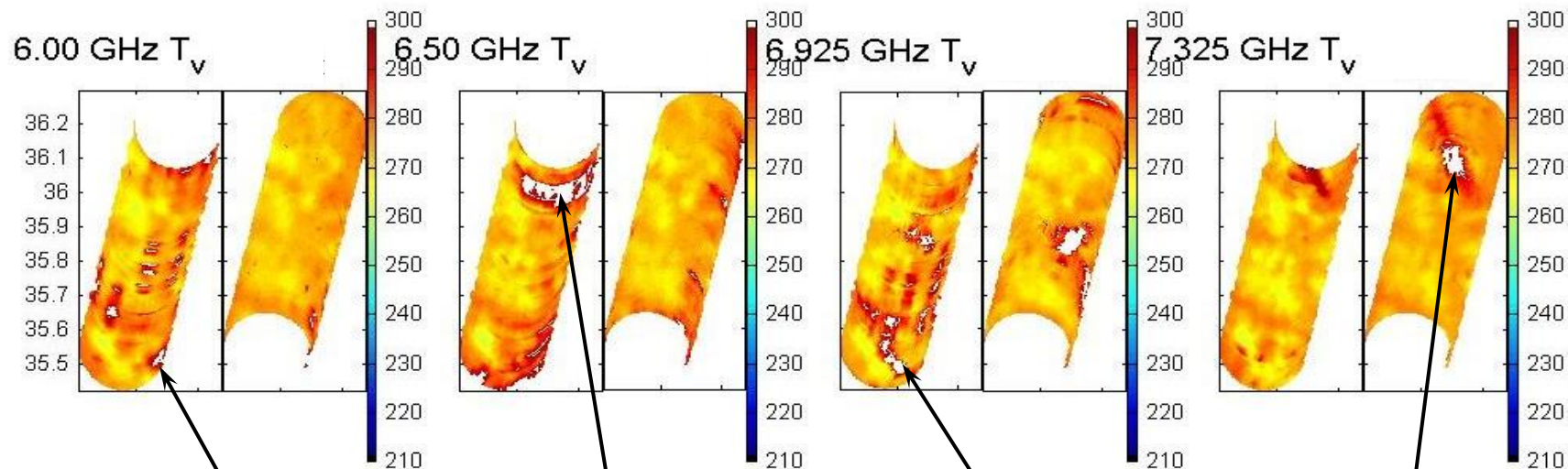
| Technique  | Use Case                                | Limitation  |
|--|---|---|
| Threshold Blanking   | General RFI                             | No Data Recovery                                      |
| Multiple Receiver Subband Mitigation (e.g. $\chi^2$ , threshold, etc.) | Narrowband RFI                          | Loss of Bandwidth                                     |
| FT Subband Threshold Blanking  | Narrowband Nyquist Sampled RFI          | Loss of Bandwidth                                     |
| Amplitude Statistic Threshold Methods (e.g. Kurtosis)                  | General Nyquist-Sampled RFI             | No Data Recovery                                      |
| FT Subband Amplitude Statistic Thresholds                              | Narrowband Nyquist-Sampled RFI          | Loss of Bandwidth                                     |
| Coherent Demodulation Bayesian Estimation                              | Nyquist-Sampled Coherent RFI            | Requires Coherence and Known Modulation & Digital RFI |
| N-Dimensional Maximum Likelihood Optimization                          | $\geq$ Nyquist-Sampled RFI w/ Knowledge | Requires Known Modulation & Digital RFI               |



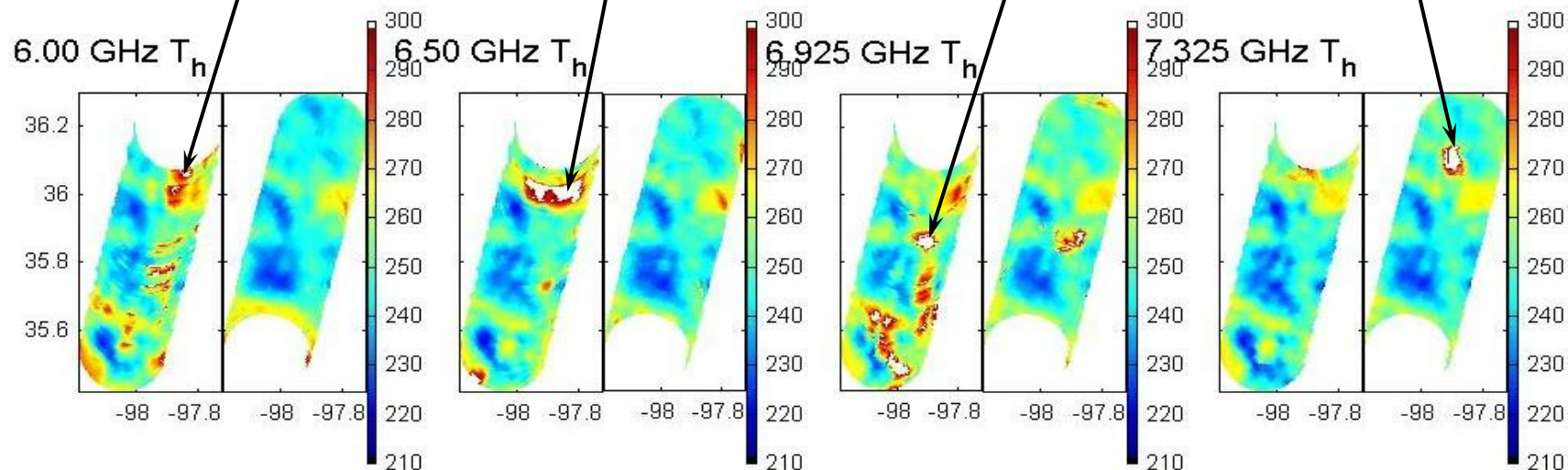
5.80-6.20 GHz (v,h)  
6.30-6.70 GHz (v,h)  
6.75-7.10 GHz (v,h,U,V)  
7.15-7.50 GHz (v,h)

# Calibrated (RFI uncorrected) Imagery

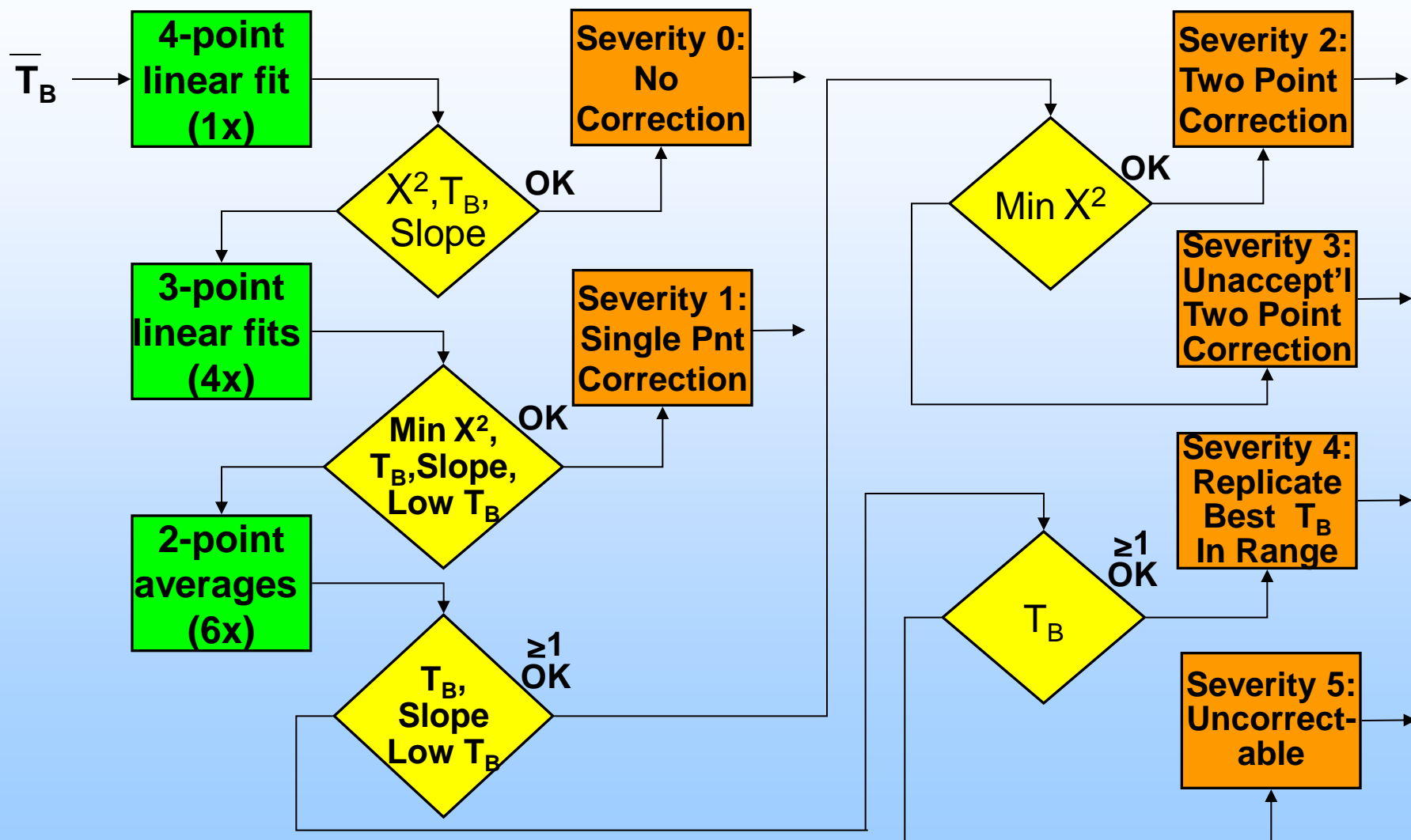
PSR/C SGP99 7/14/99 – Oklahoma – SN 0049



**Interference above geophysical and instrument noise from ground-based active services**



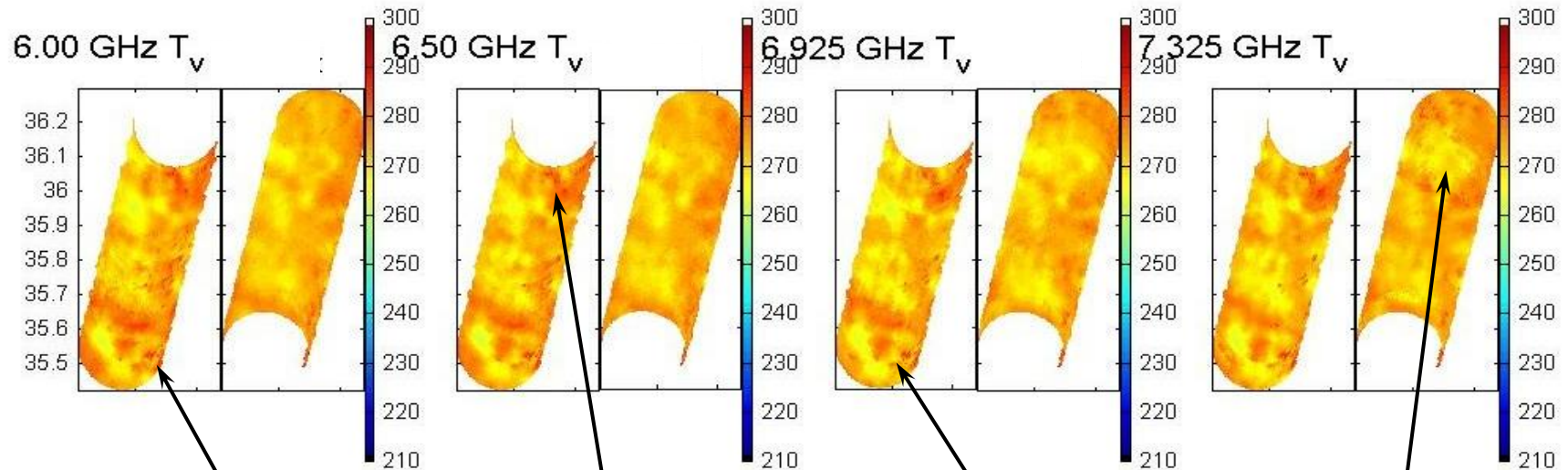
# Basic Spectral Consistency Algorithm



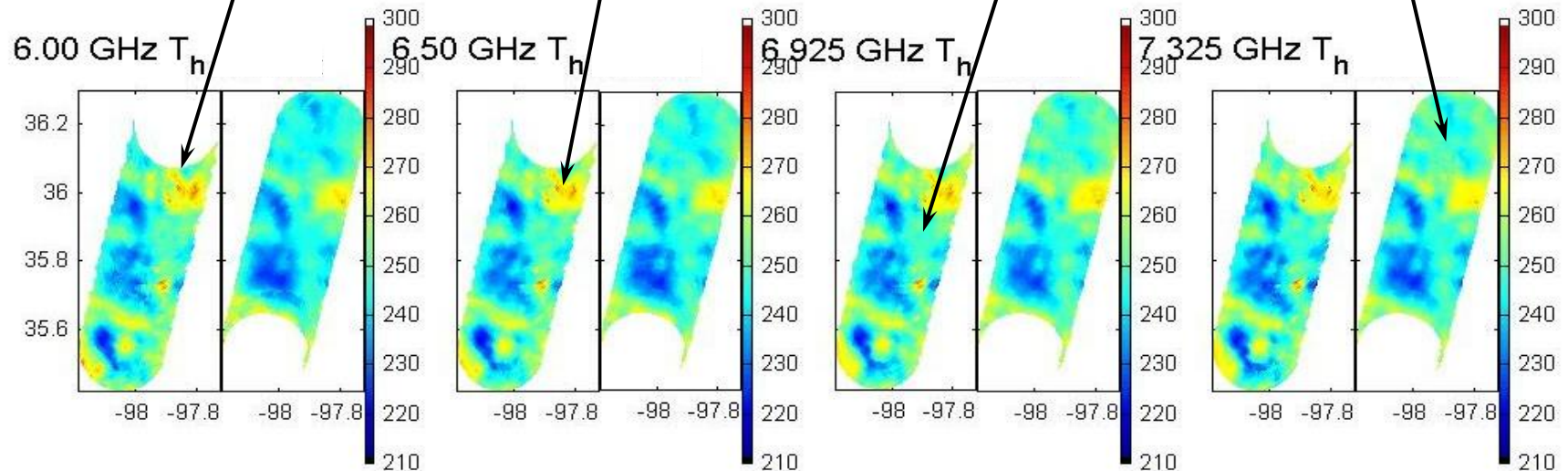


# Interference-Corrected Imagery

PSR/C SGP99 7/14/99 – Oklahoma – SN 0049



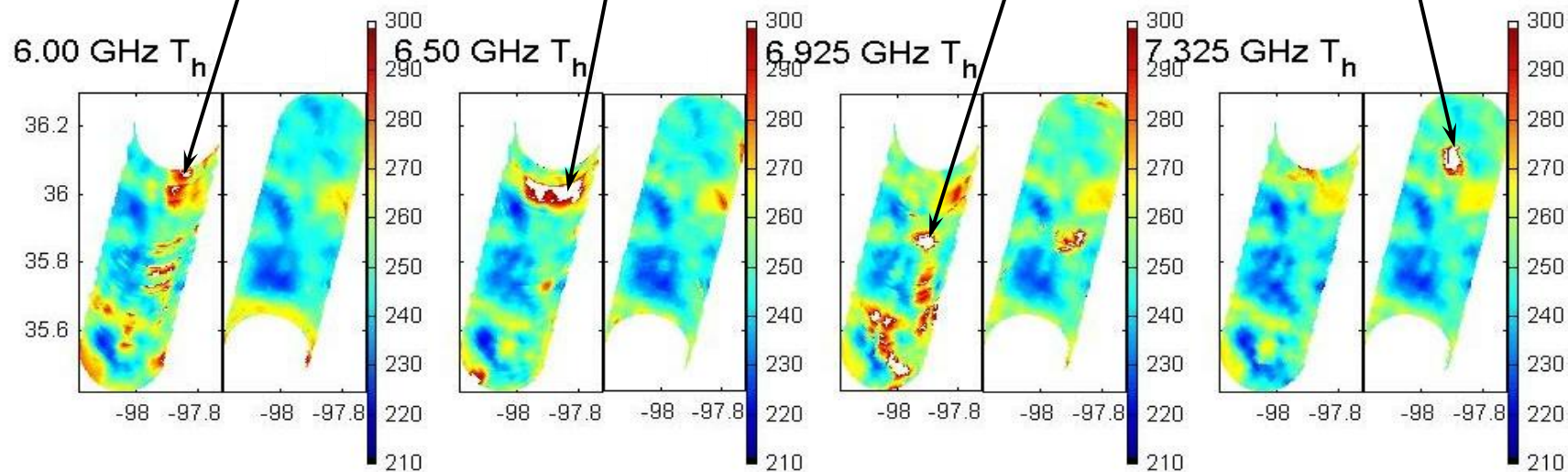
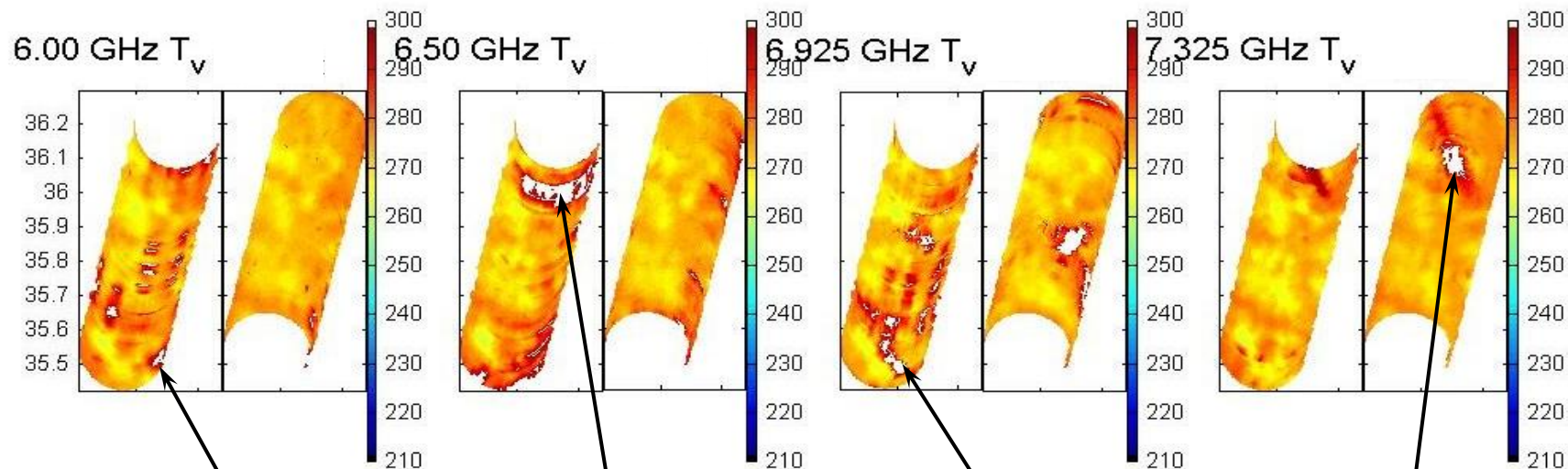
**Interference mostly removed for purposes of soil moisture measurement**





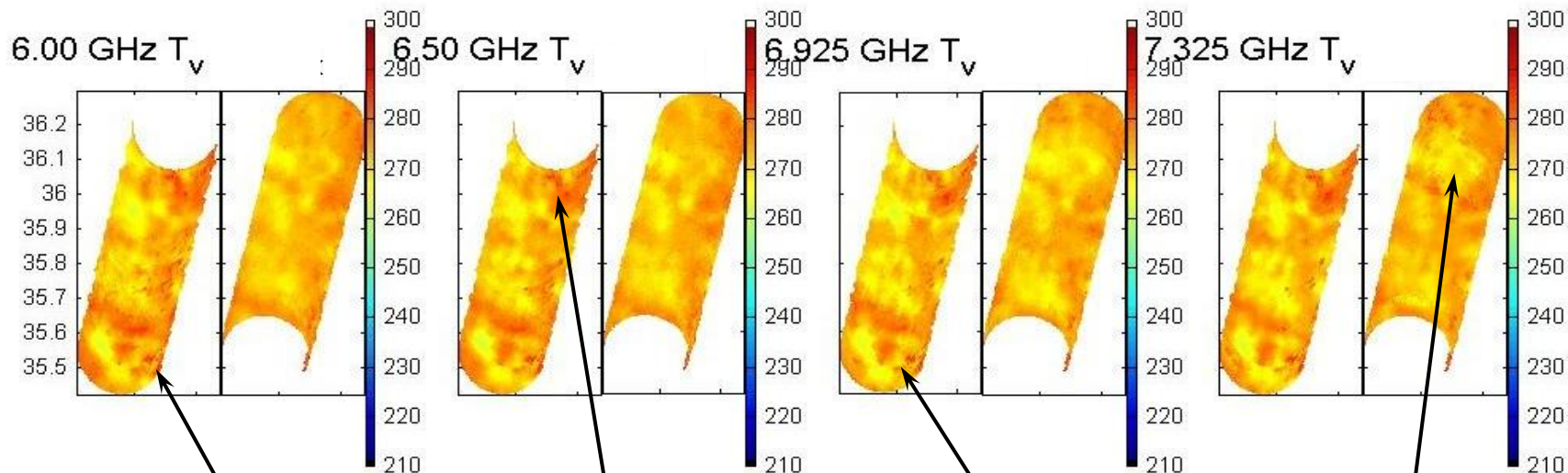
# Calibrated (RFI uncorrected) Imagery

PSR/C SGP99 7/14/99 – Oklahoma – SN 0049

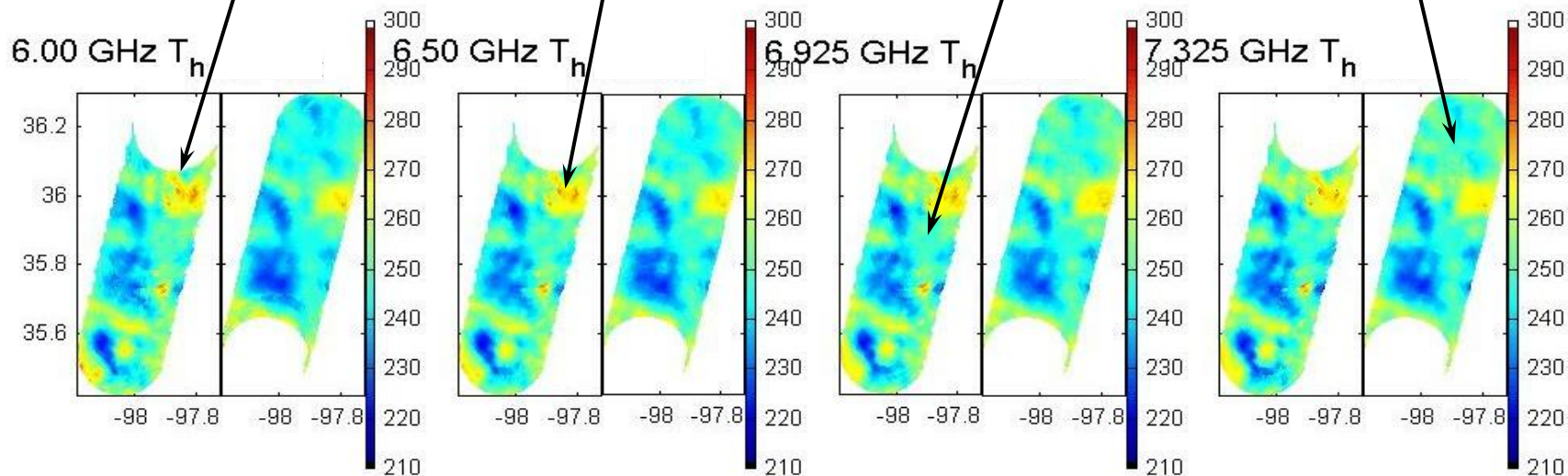


# Interference-Corrected Imagery

PSR/C SGP99 7/14/99 – Oklahoma – SN 0049



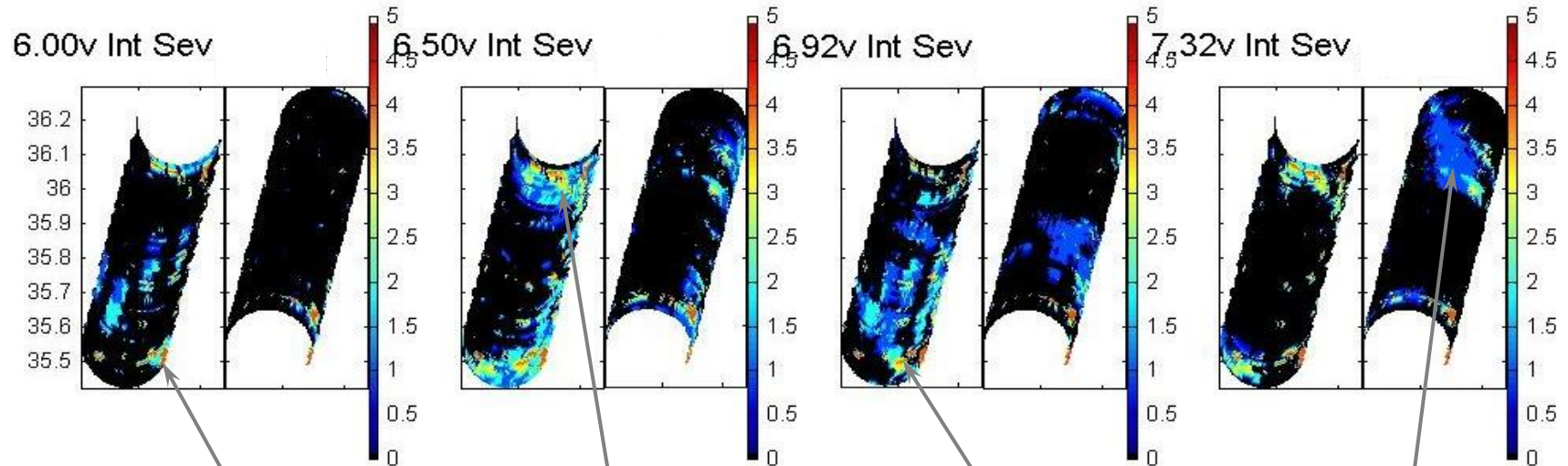
**Interference mostly removed for purposes of soil moisture measurement**



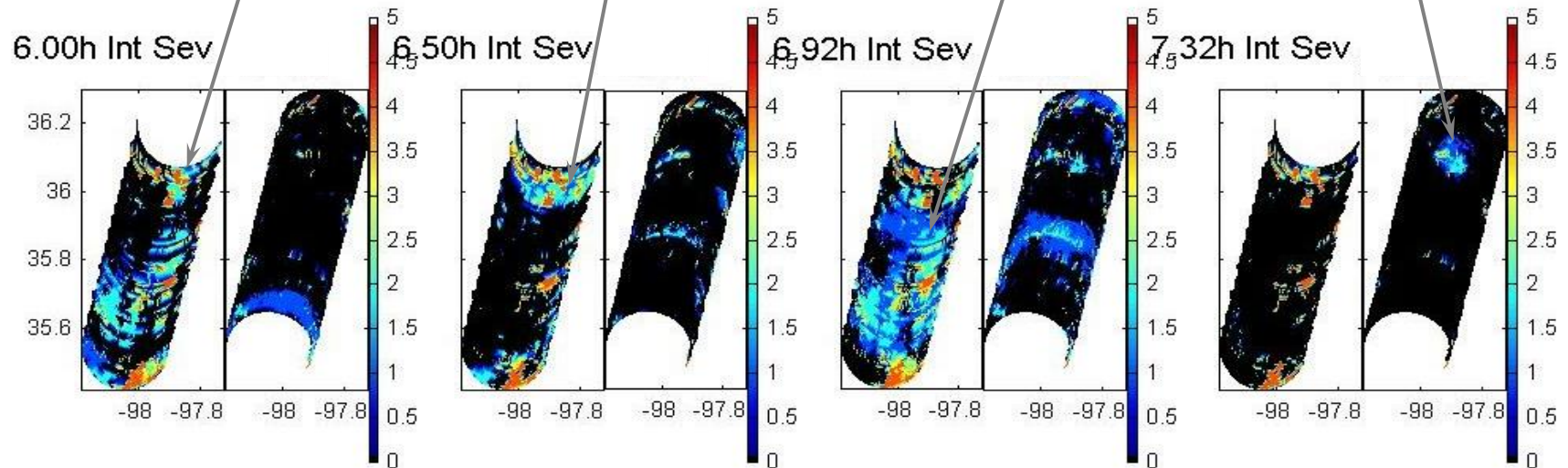


# Interference Severity Maps

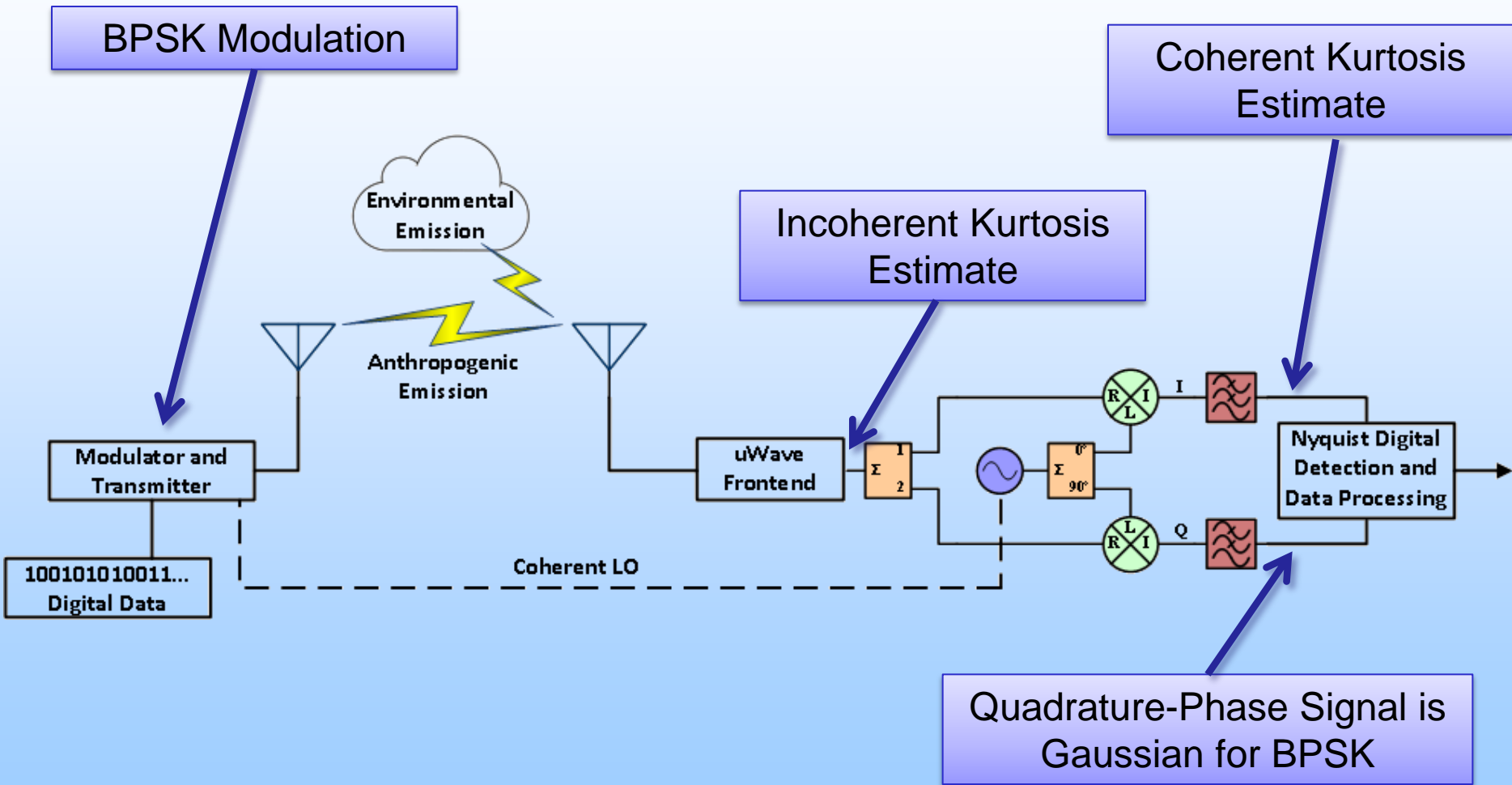
PSR/C SGP99 7/14/99 – Oklahoma – SN 0049



**Interference severity varies according to subband, look direction, polarization & location**

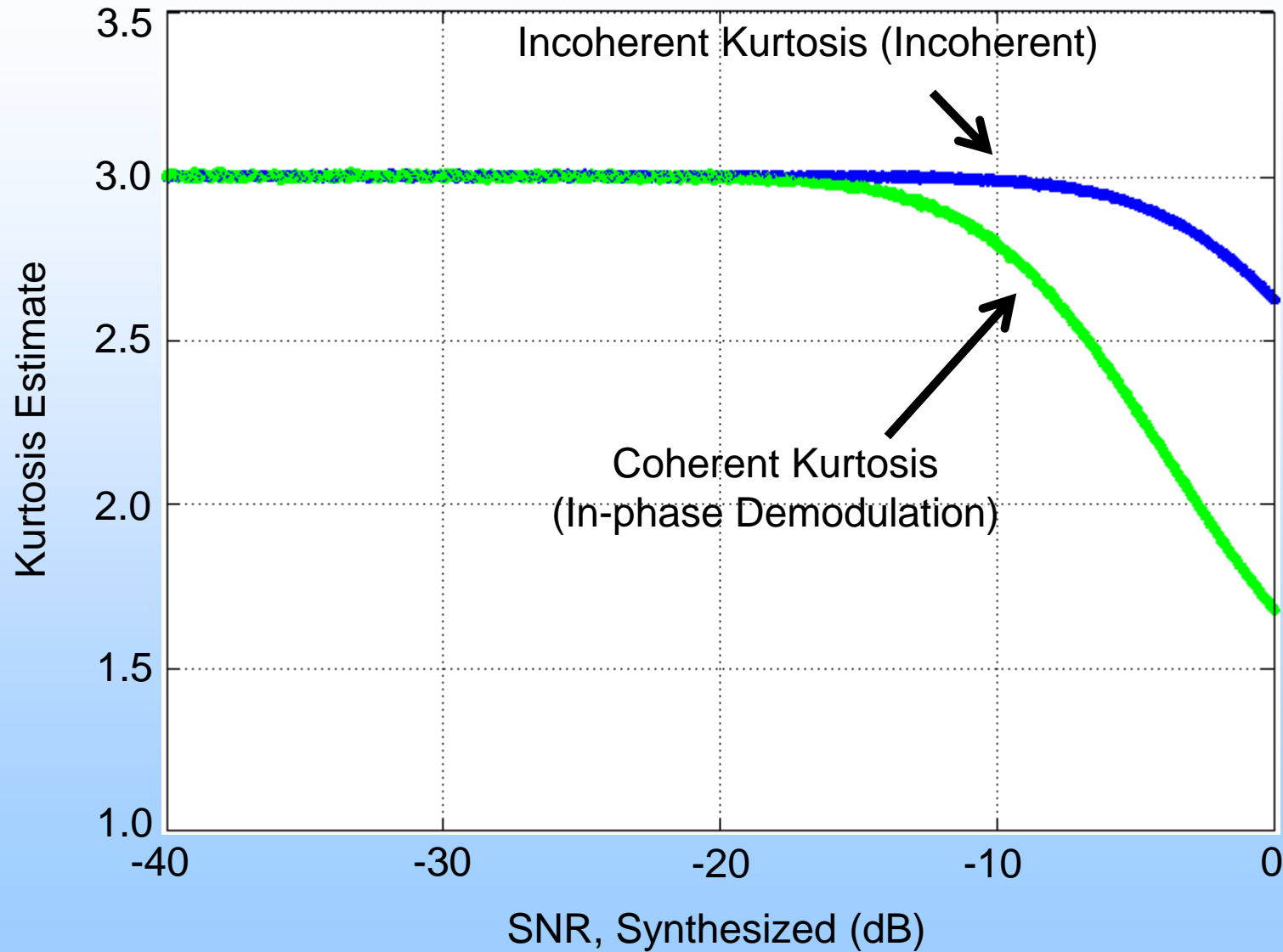


# BPSK + Thermal Noise Simulation



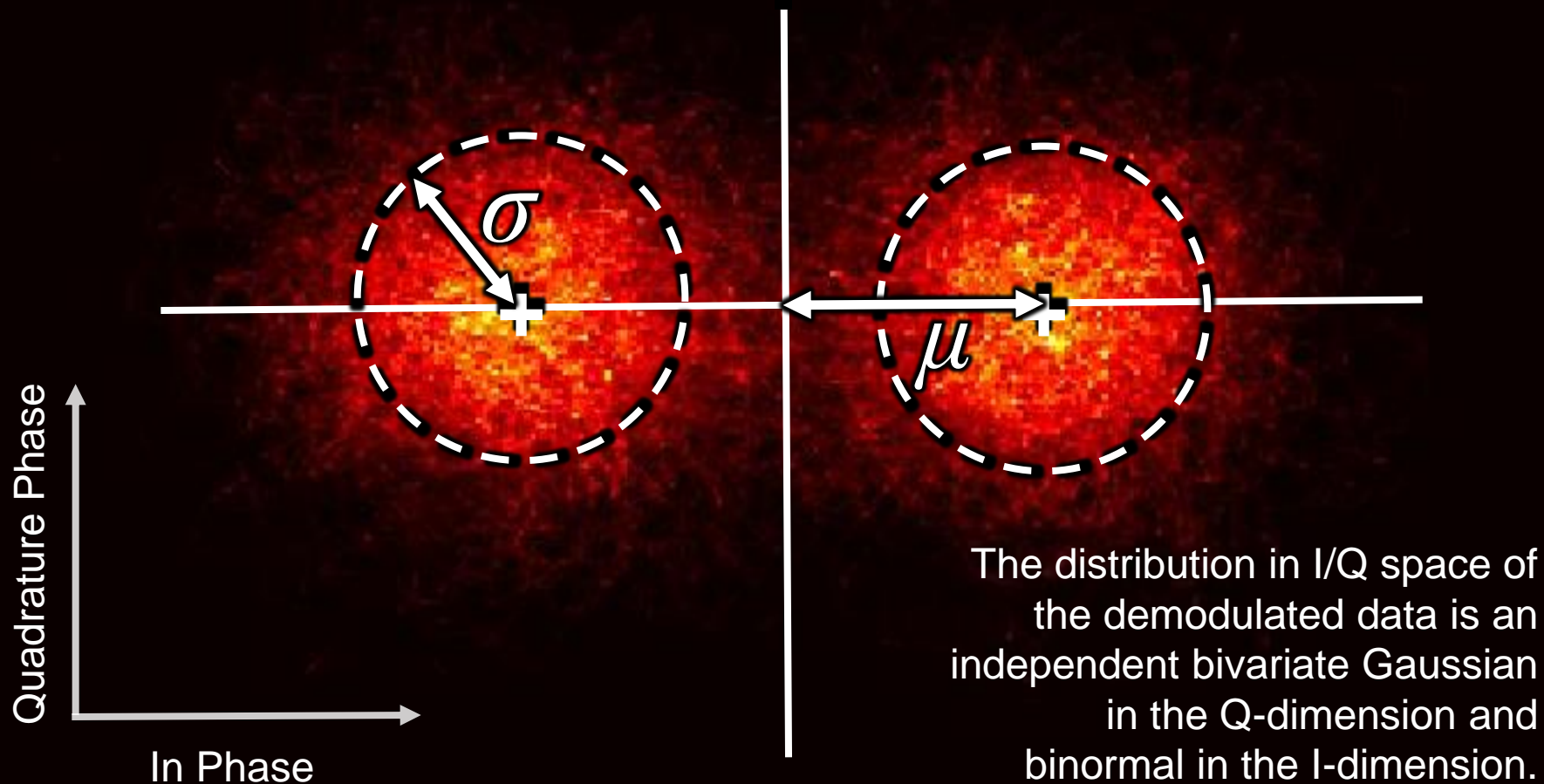


# Kurtosis RFI Mitigation

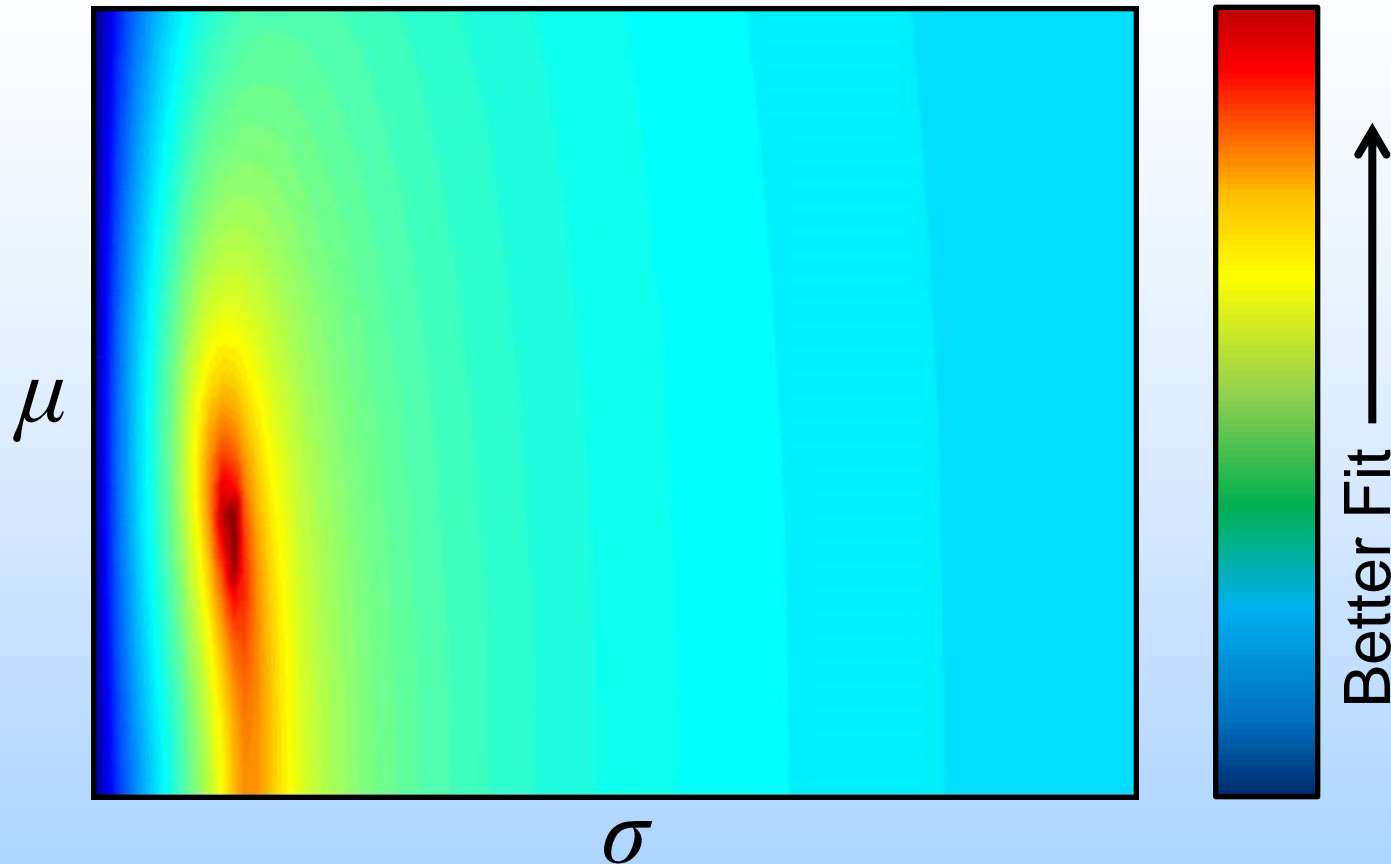


$$SNR \text{ dB} = 10 \cdot \log \left( \frac{\text{RFI Signal Power}}{\text{Environmental Noise Power}} \right)$$

e.g., SNR: -2.54 dB

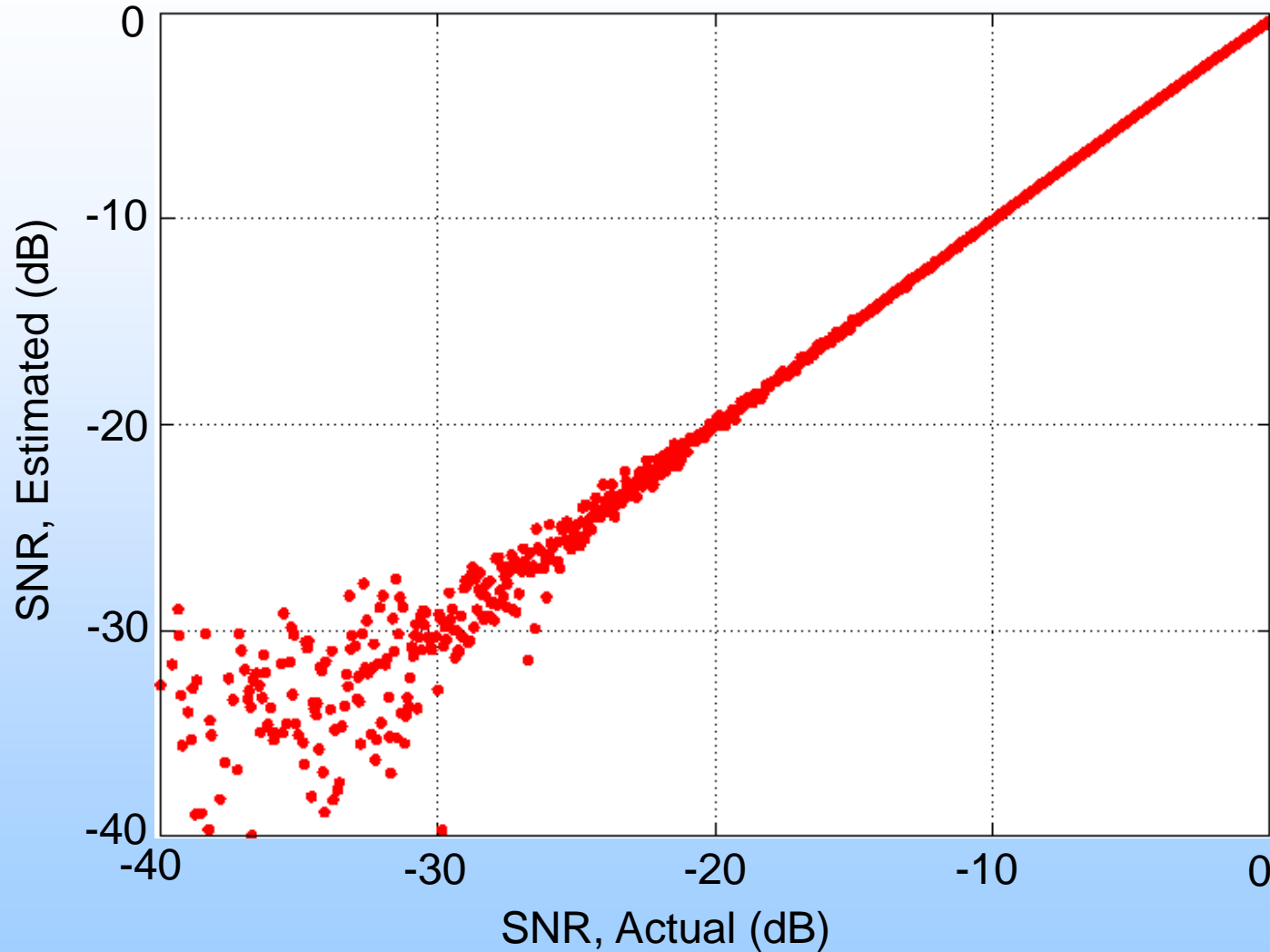


# Maximum Likelihood PDF Parameter Estimation



Likelihood function based on Nyquist sampled data is well behaved with a single unambiguous maximum. Requires I/Q sampled vector or additional LO phase estimation.

# ML SNR Estimation



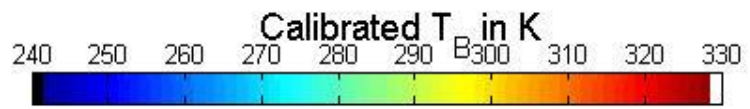
1000 model runs between -40 and 0 dB SNR



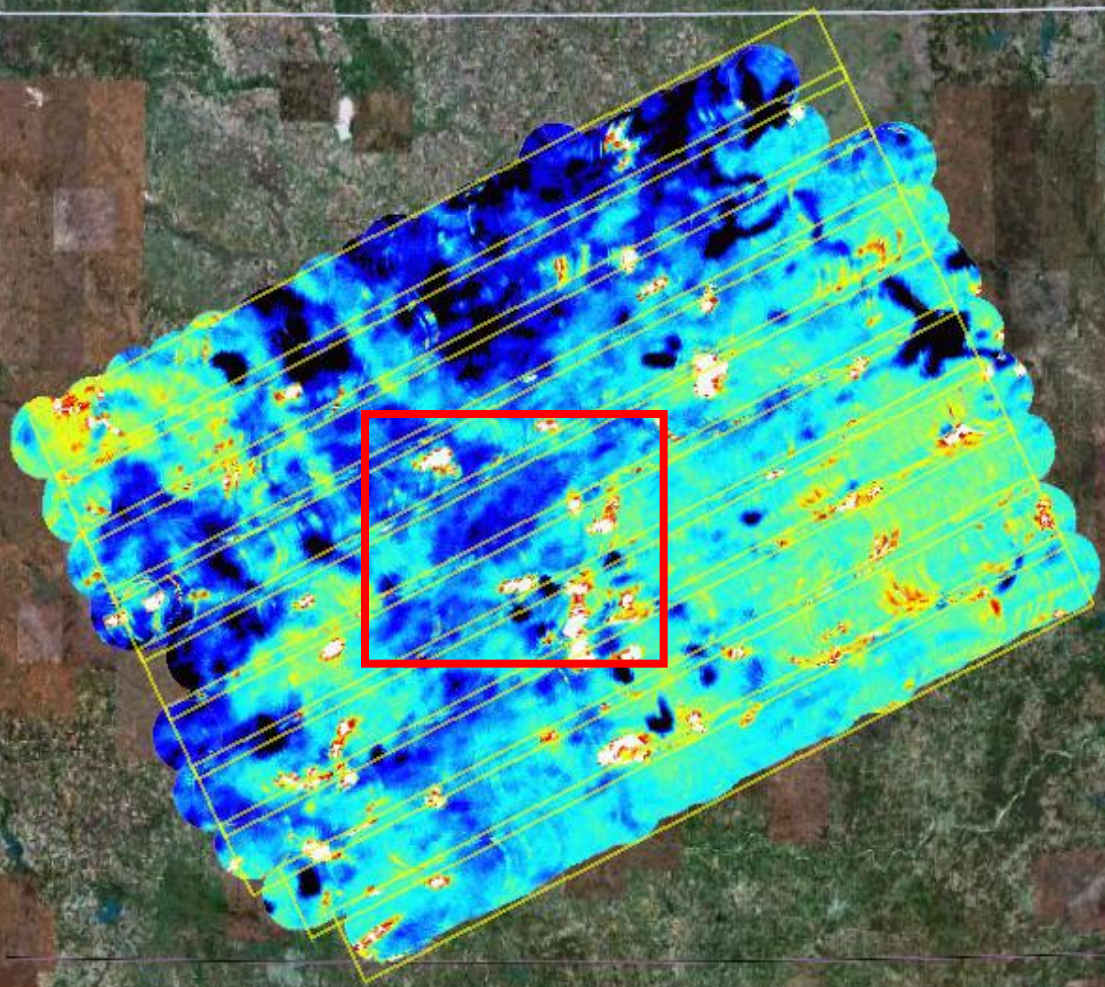
Airborne imaging systems with ***efficient*** ( $\eta_t \sim 1$ ), ***sensitive*** ( $\Delta T_{\text{RMS}} \sim 0.5\text{-}1\text{K} \leftrightarrow k(\delta T)B \sim -128\text{ dBm in } 10\text{ MHz}$ ), and ***calibrated*** ( $\Delta T_{\text{BIAS}} \sim 0.5\text{-}1\text{K}$ ), radiometers are an effective (\$ cost, time, area coverage) means of locating interference.

- Proposed to OSTP as a recommendation from the NRC spectrum study in 2012 (~\$10M)
- Demonstrated by CET instruments during over 1200 flight hours since 1996 on manned aircraft (but at high \$ cost)
- Now being implemented on small UAS (at low \$cost) for commercially-driven applications (e.g., precision agriculture)





6.92 GHz H-pol



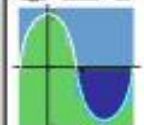
89 mi

Image State of Arkansas  
© 2007 Europa Technologies  
Image © 2007 TerraMetrics

© 2007 Google

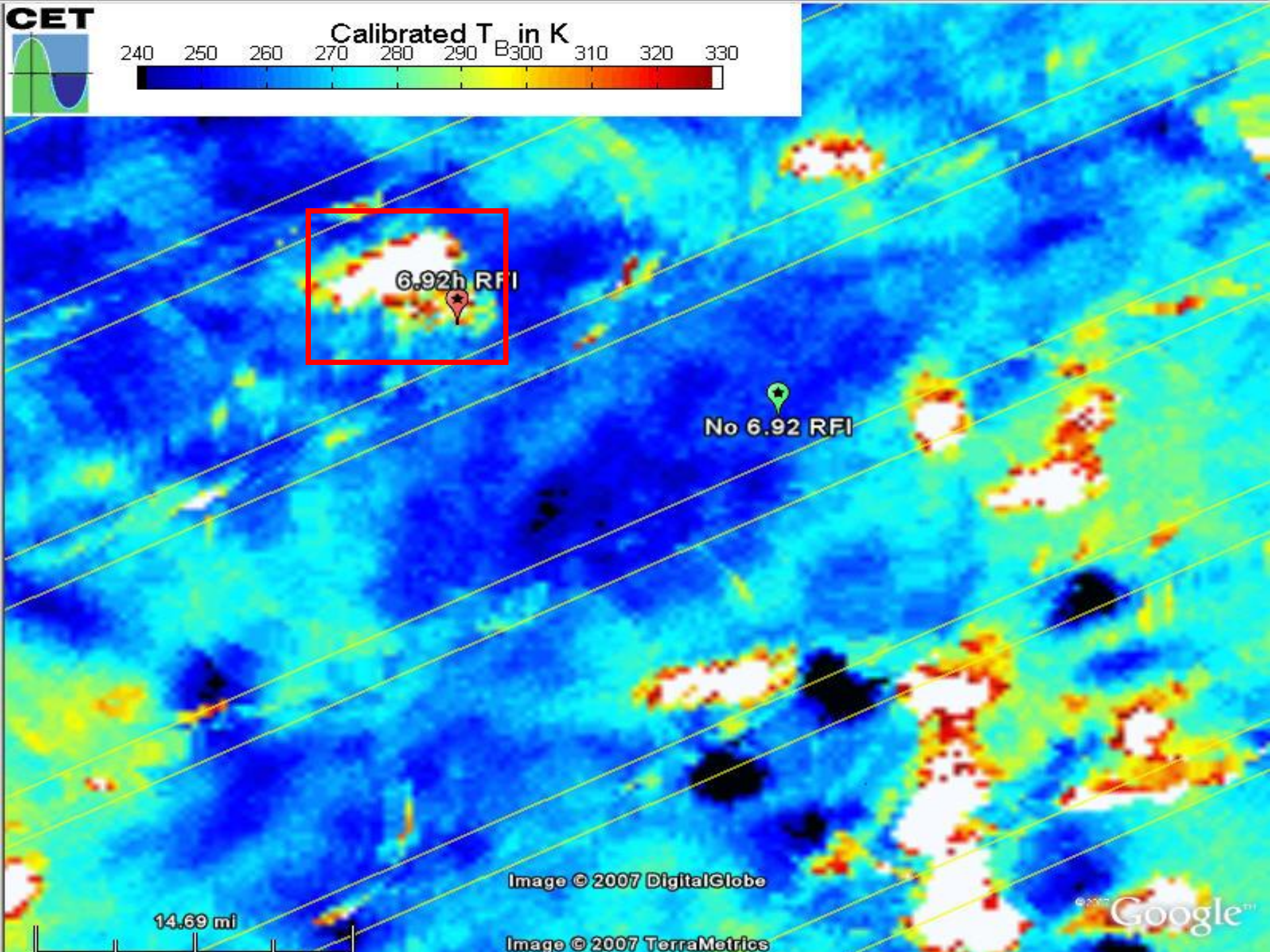


CET



Calibrated  $T_{B300}$  in K

240 250 260 270 280 290 300 310 320 330



6.92h RFI

No 6.92 RFI

Image © 2007 DigitalGlobe

Image © 2007 TerraMetrics

14.69 mi

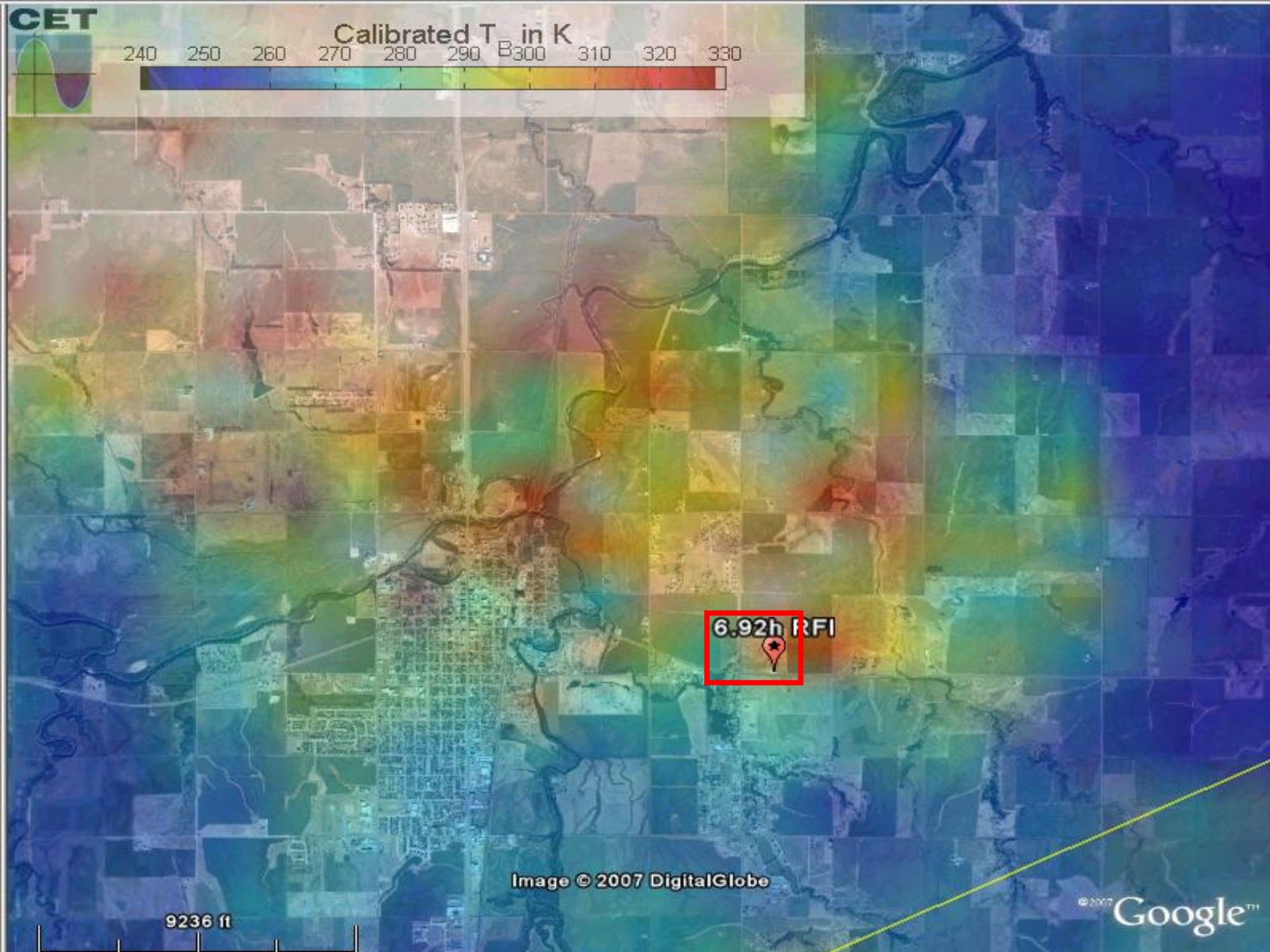
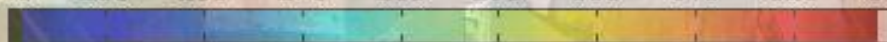
Google™



CET

Calibrated  $T_{B300}$  in K

240 250 260 270 280 290 300 310 320 330



6.92h RFI

Image © 2007 DigitalGlobe

9236 ft

© 2007 Google™



CET



Calibrated  $T_b$  in K

240 250 260 270 280 290 300 310 320 330



6.92h RFI

CE F. 98000 NOE

Image © 2007 DigitalGlobe

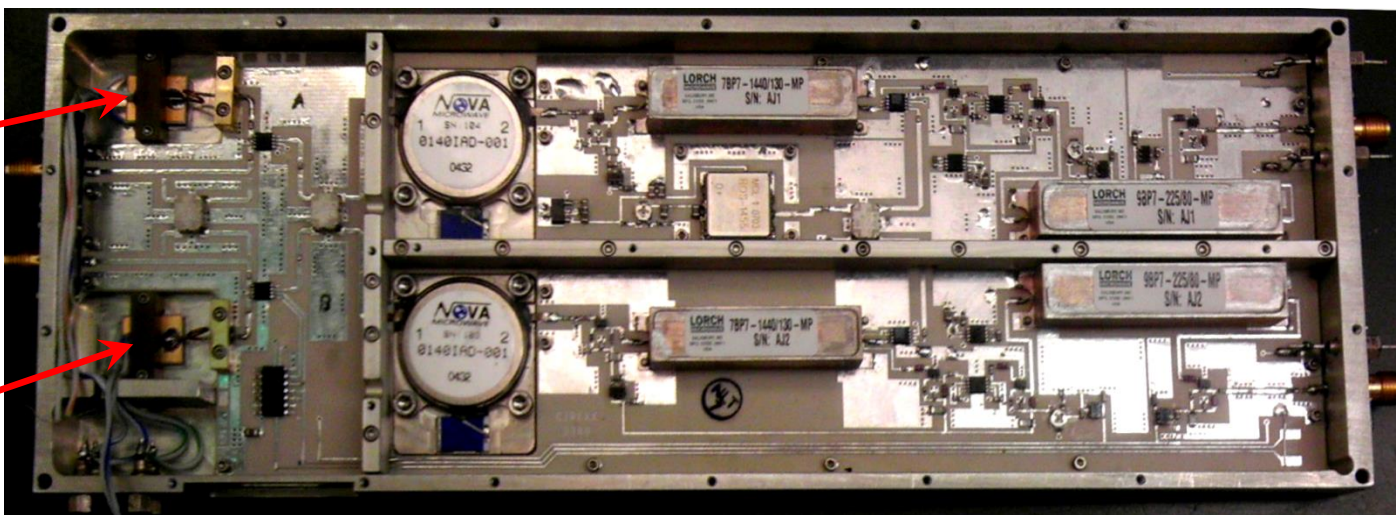
375 ft

© 2007 Google™



# LDCR Receiver (2015 Build)

Calibration  
terminations  
w/ TE  
cooler/  
heaters

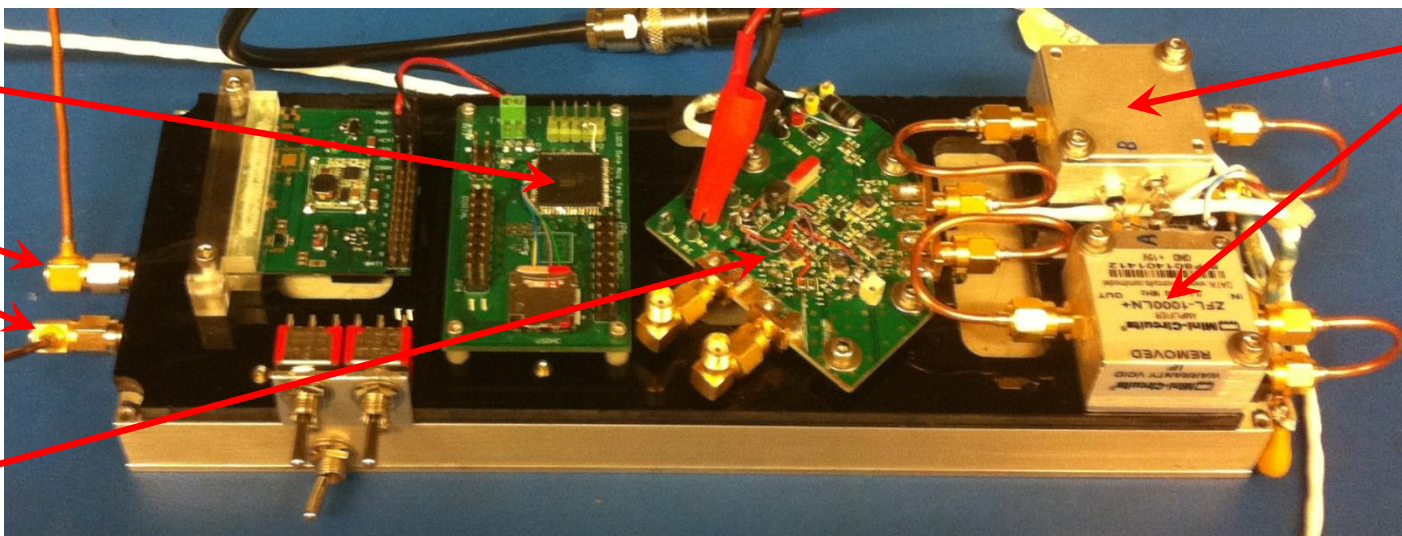


uWave  
receiver+  
carrier  
mass:  
550 g

Micro-  
controller  
w/ ADC

Up/down  
RF inputs

Analog  
correlator/  
video amp



IF gain

Total  
mass:  
881 g

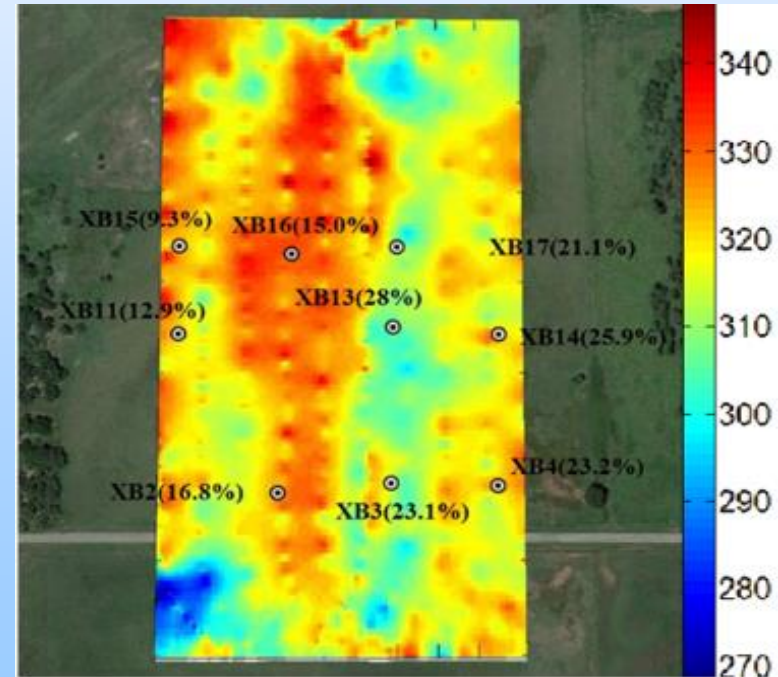
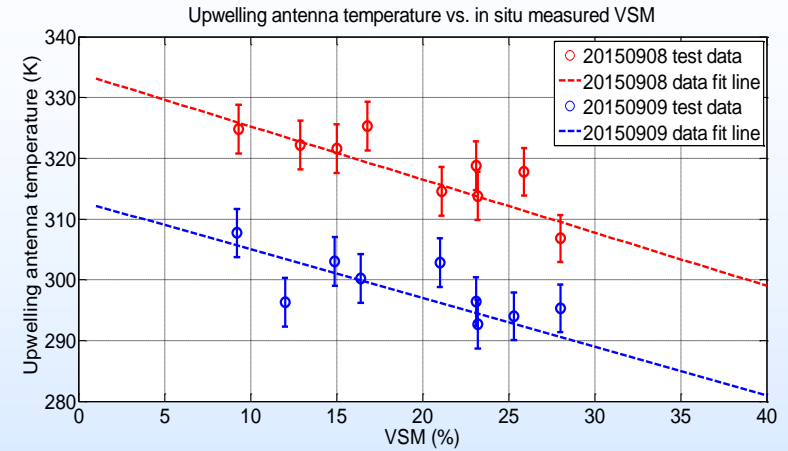
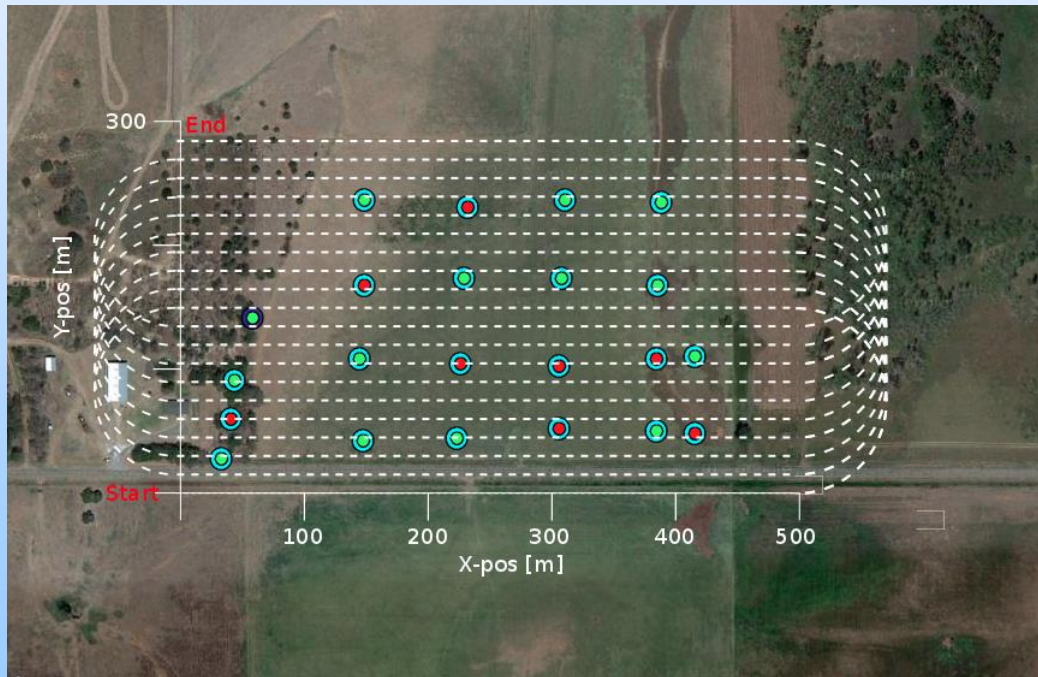
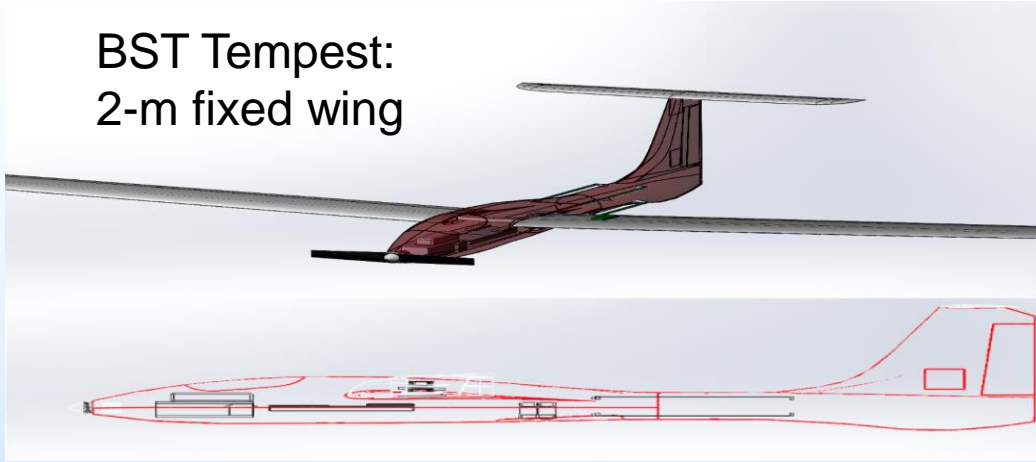
(battery  
add'l  
364 g)

LDCR prototype receiver (23x8x4 cm) with analog correlator, A/D, and microcontroller.



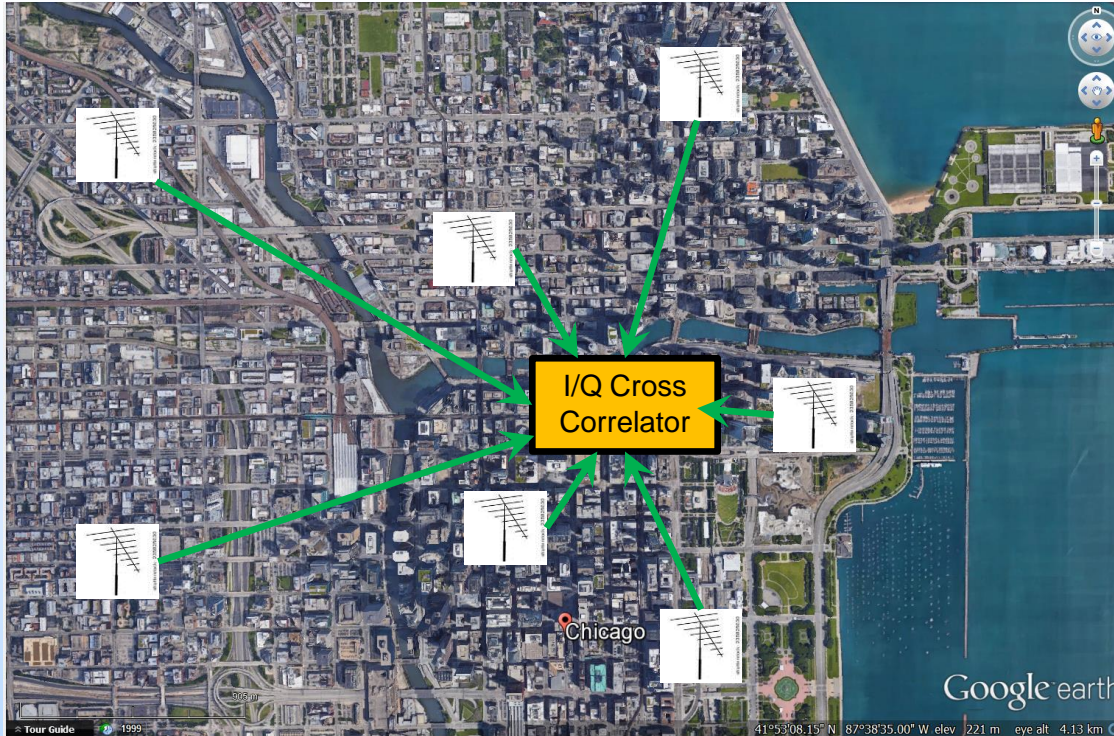
# L-Band Soil Moisture Mapping using Small UAS

BST Tempest:  
2-m fixed wing



# Coherent Spectrum Monitoring

## Persistent “Inverse GPS”



$$\Delta r \sim c/\Delta f$$

$$T_s \sim 2/\Delta f$$

$$N_s \sim R_{max}/(cT_s)$$

### Example:

$$\Delta f = 100 \text{ MHz}$$

$$\Delta r \sim 3 \text{ m}$$

$$T_s \sim 200 \text{ MSps}$$

$$R_{max} = 45 \text{ km}$$

$$N_s \sim 2^{15}$$

- Nyquist-sampled data from several widely spaced receivers can be time-correlated to geolocate sources (uses phase/time delay and amplitude)
- Transient and multiple sources near noise floor can potentially be accurately geolocated and time-tagged. Use a-priori knowledge to narrow search/track.
- Requires central (not distributed) wideband FPGA-based cross-correlation processor and wideband (e.g., fiber optic or optical free space) data pipes (~10 Gbps).





# Manifold-Based System for Passive-Active Spectrum Sharing

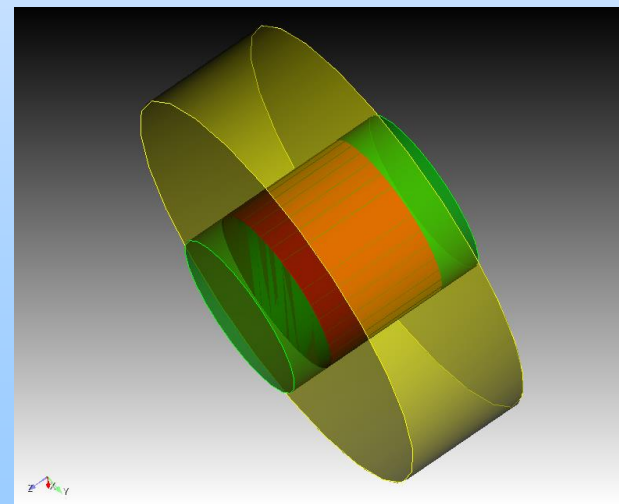
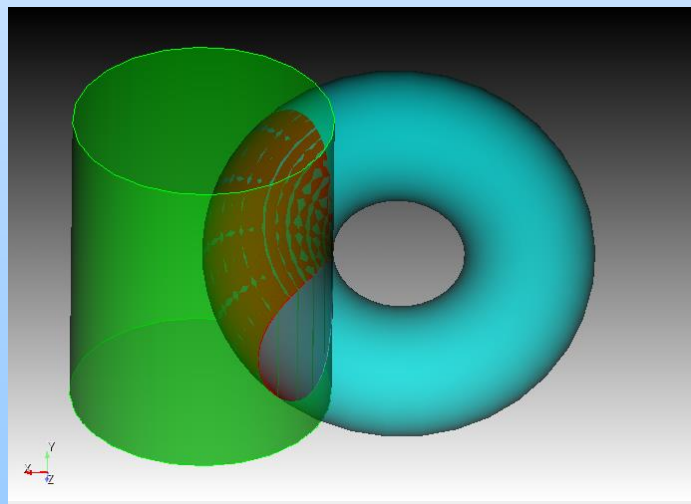


**Objective:** To develop an automated scalable technique for identifying interference in 7-dimensional “electrospace” ( $f, x, y, z, \theta, \phi, t$ ).

The spectral, spatial, angular, and temporal needs of each eligible service within a band can be defined by a **manifold** (i.e., hypervolume) in 7-D electrospace (Joshua Chong, MS Thesis, 2014).

Intersections of manifolds identify **competitive hypervolumes** for which priority-based arbitration or interference flagging can be used.

Competitive volumes  
in 3-D space  
illustrated  
by the intersection  
of one or more 2-D  
manifolds (i.e.,  
differentiable  
surfaces)



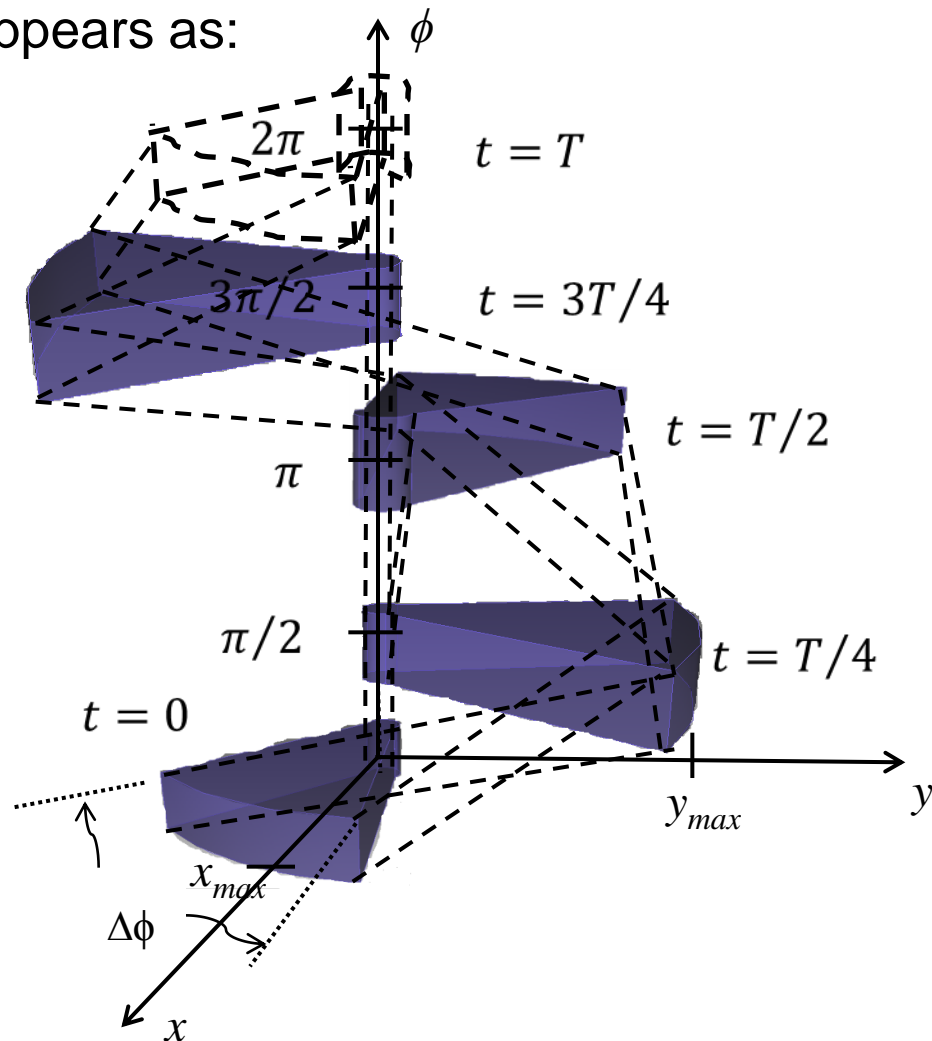
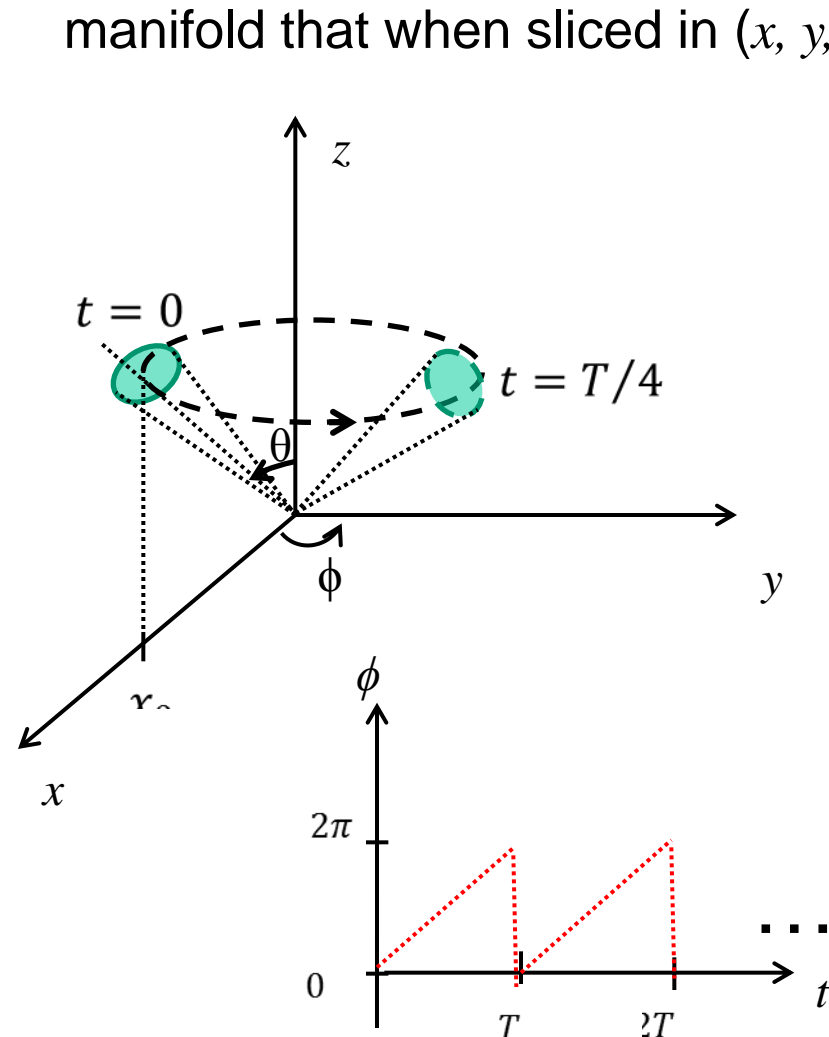


# Visualization of Manifolds

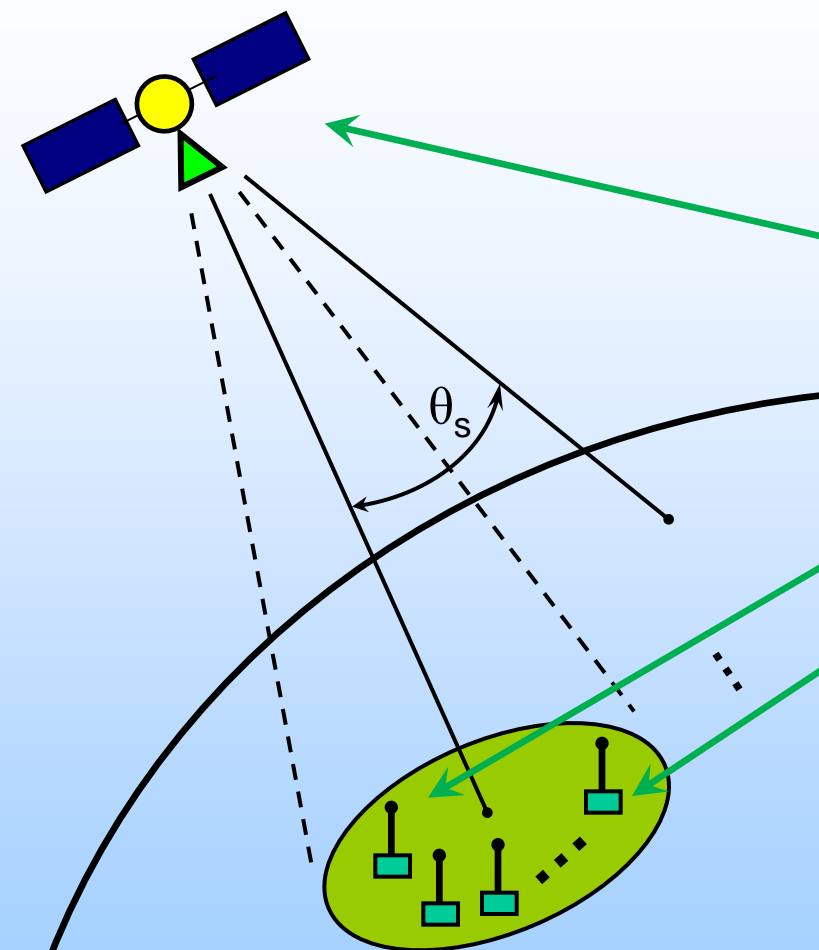
## Ground-Based Terminal Area Radar



Manifolds can be depicted as **time-dependent hypervolumes** sliced in 3-D space. A **conically-scanned terminal area radar** uses a transmit manifold that when sliced in  $(x, y, \phi)$  appears as:



# Manifold Based RFI Mitigation



- Users signal their intention to use electrospace on ~1-10 msec intervals
- Spectrum “broker” provides a real-time non-compulsory assessment of RFI
- Users are free to act on RFI data as they see fit

- Provide 3-5 year graduate fellowships. *This is an investment opportunity!*
- Provide year-round opportunities for small and regular hardware grant proposals
- Support regular forums for research dissemination (URSI NRSM Commission E sessions)
- Good example: NSF EARS (but opportunities are not often enough, too much “winner-take-all”)



**AUGUST 3<sup>rd</sup>**

**2:45 Panel**

**Data Analytics**

**Moderator: Tim Hall, NIST/CTL**

This panel will focus on the mathematical techniques for analyzing captured signal data, with an emphasis on machine learning, data mining and artificial intelligence techniques. Classical pattern recognition and other statistically-based methods may also be included. The limitations and potential pitfalls of each technique will be considered along with the advantages and capabilities. Related to this topic is the question of what type and quantity of data need to be captured and retained in order to use each technique. Time and frequency resolution and appropriate quantization are examples of this. The panel will not only discuss current research results but will help identify specific topics and techniques where future research is still needed.

**Panelists:**

- **Tim O'Shea – Virginia Tech**
- **Tom Rondeau – DARPA**
- **Al Gasiewski – Science Band, University of Colorado**
- **Mitch Kokar – Northeastern University**
- **Dirk Grunwald – University of Colorado**

**4:45 Closing Remarks Keith Gremban, NTIA/ITS.D**