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# medNoiseChar: Assessing LIGO Data Stationarity

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# Introduction

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- Problem:
  - » Well-behaved detector has stationary noise, calibration
  - » Data analysis methods rely on stationarity of noise, calibration
  - » How to assess, monitor calibration, noise stationarity?
- medNoiseChar
  - » Monitors variation of three measures of detector noise statistics and identifies approximately stationary noise segments
- Outline
  - » Characterizing detector noise
  - » Monitoring stationarity
  - » Identifying stationary epochs
  - » A look at S2 stationarity



# Basic Idea

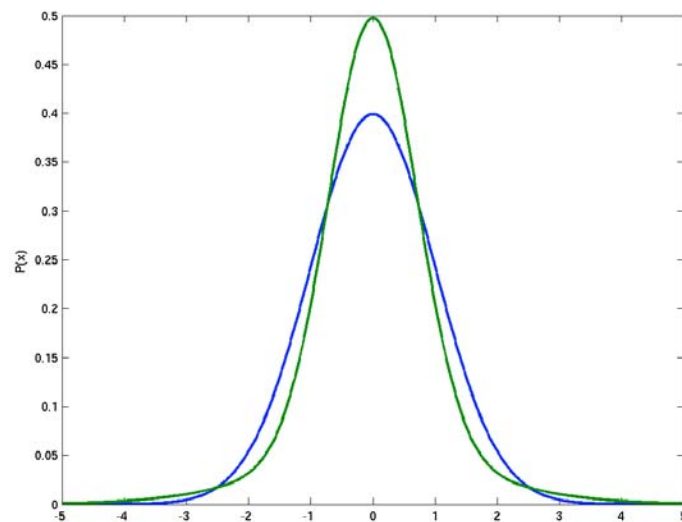
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- Identify interesting, relevant measures of detector noise statistics
  - » E.g., mean, variance in particular band
- Evaluate statistic periodically
  - » E.g., on consecutive 16s intervals
- Is measurement on current segment consistent with measurement on previous segment? With measurement on past several segments?
  - » Each measurement is an estimate: consistency is a test based on distribution of errors associated with each measurement
- Stationary epoch: longest set of consecutive intervals whose statistics are “consecutively consistent”



# Characterizing detector noise

- Central tendency
  - » E.g., mean
- Dispersion
  - » E.g., variance
- “Gaussianity”
  - » E.g.,  $\chi^2$  fit to Gaussian
- Problem: Detector noise shows strong, rapid variation in sample distribution wings
  - » Timescale: seconds or less
  - » Wings:  $2-3\sigma$
  - » Variability: strong, asymmetric departures from Gaussianity apparent in mean, dominate variance

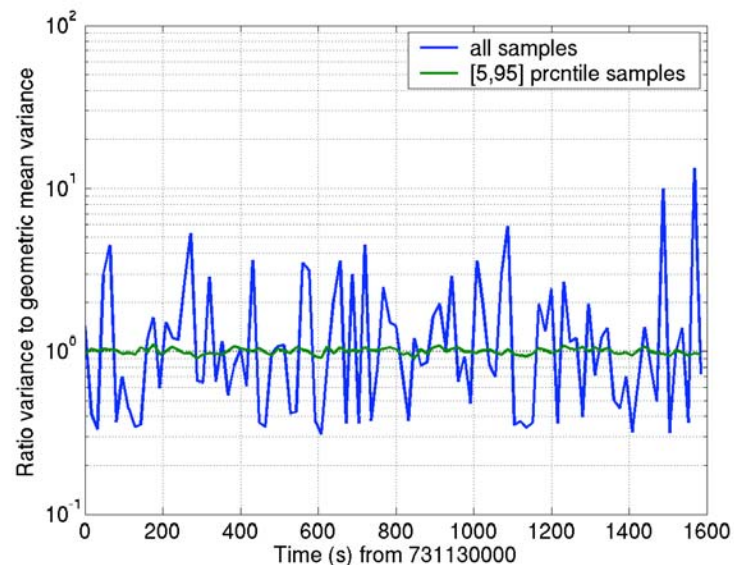


Same mean, variance; different  $\chi^2$  fit to Gaussian



# Characterizing detector noise

- Focus on distribution mass, not outliers
  - » Rank order N samples per averaging time
  - » Drop first, last pN samples
    - These are the outliers
  - » Evaluate mean, variance,  $\chi^2$  fit to Gaussian of remaining samples
  - » Measures are resistant to changes in outliers
- [5,95] percentile cut
  - » Reduces variability in dispersion on 16 s intervals by > factor 30 on typical segment
- *Caveat: musn't ignore outlier variability*
  - » But that's a different study



	$\sigma_v / \langle \sigma_v \rangle$	$\sigma_v / \langle \sigma_v \rangle$ [5,95]
Gaussian noise	0.0063	0.0063
S2 data	1.8823	0.0414

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# Monitoring stationarity

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- Focus on properties of truncated distribution estimated on consecutive data segments
  - » E.g., consecutive 16s intervals
- Mean
  - » N samples: error bars (approximately) proportional to standard error of mean ( $\sigma/N^{1/2}$ )
  - » Compare to last interval mean, accumulated mean since epoch start
- Variance
  - » Ratio of two variance estimates from same distribution follow  $f$ -distribution (depends on number of samples in each estimate)
  - » Compare current variance estimate to last estimate, accumulated estimate since epoch start
- “Gaussianity”
  - » Ratio of two  $\chi^2$  quality of fit to model distribution follow  $f$ -distribution (depends on number samples in each estimate)
  - » Compare current fit quality to last



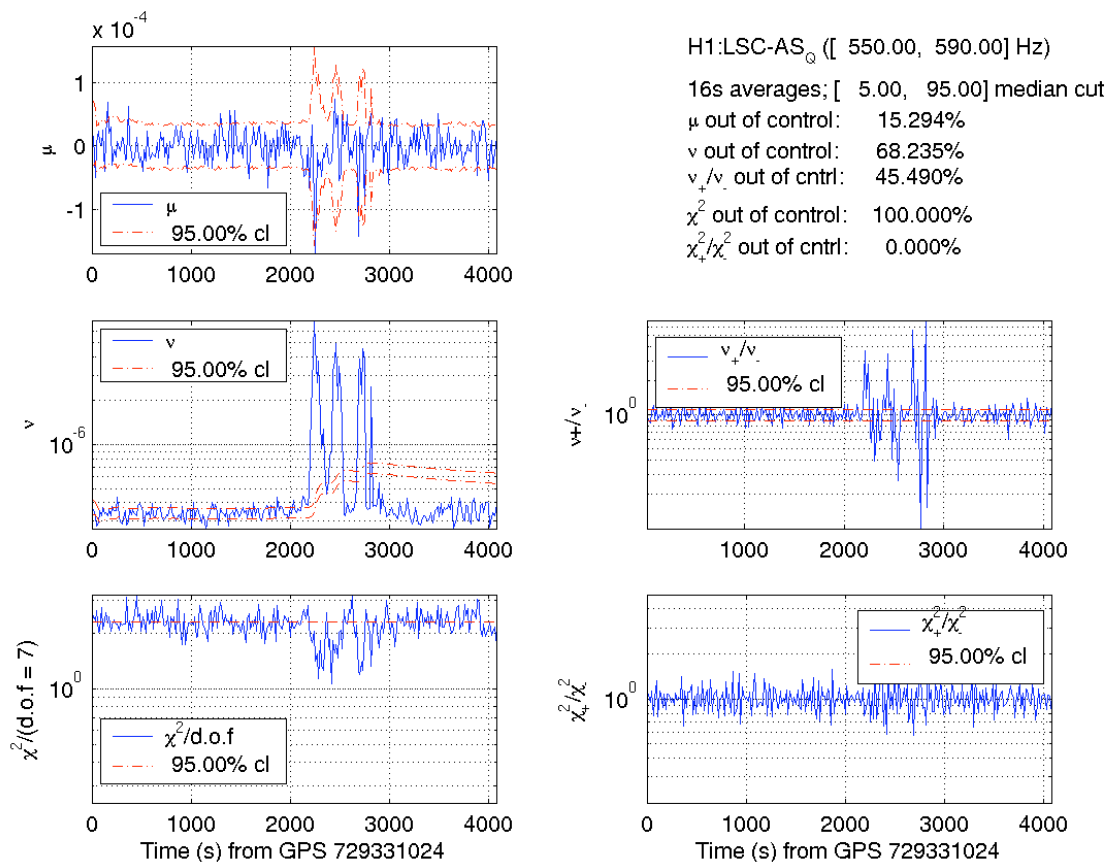
# Choices

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- Band
  - » Focus attention on band  $[f_l, f_h]$
- Averaging time
  - » Interval duration  $\tau$  on which statistics are accumulated
- Percentile cut
  - » Define bulk distribution by throwing out top, bottom  $p$  percentile of band-limited samples
- Number of bins for  $\chi^2$  fit to Gaussian
  - » Bins chosen for constant probability per bin
- Stationarity criteria
  - » Stationary epoch ends immediately before first of two consecutive intervals that are inconsistent (probability  $p$ ) in either mean or variance with epoch cumulative mean, variance



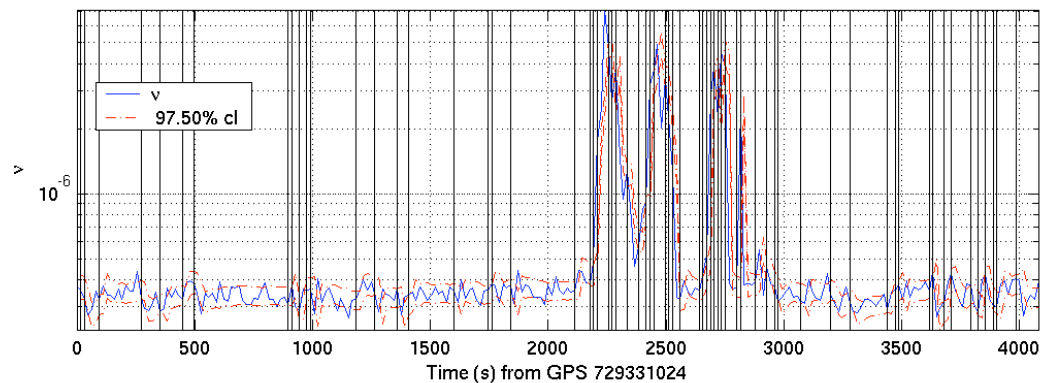
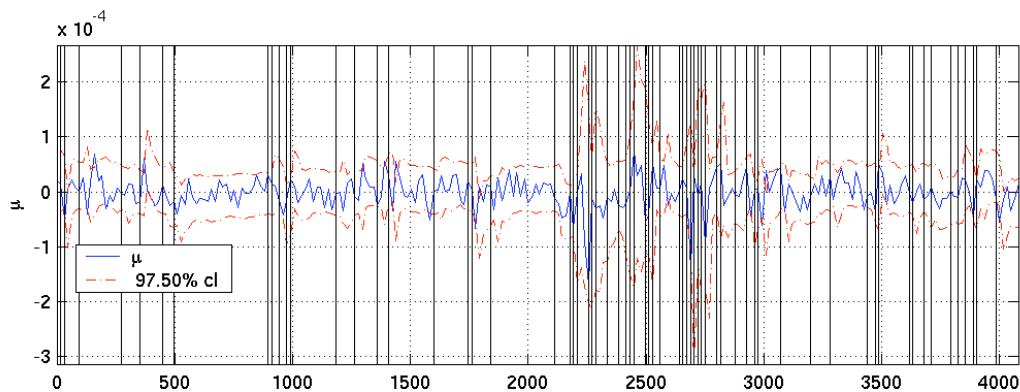
# Sample output: monitoring stationarity







# Sample output: stationary epochs



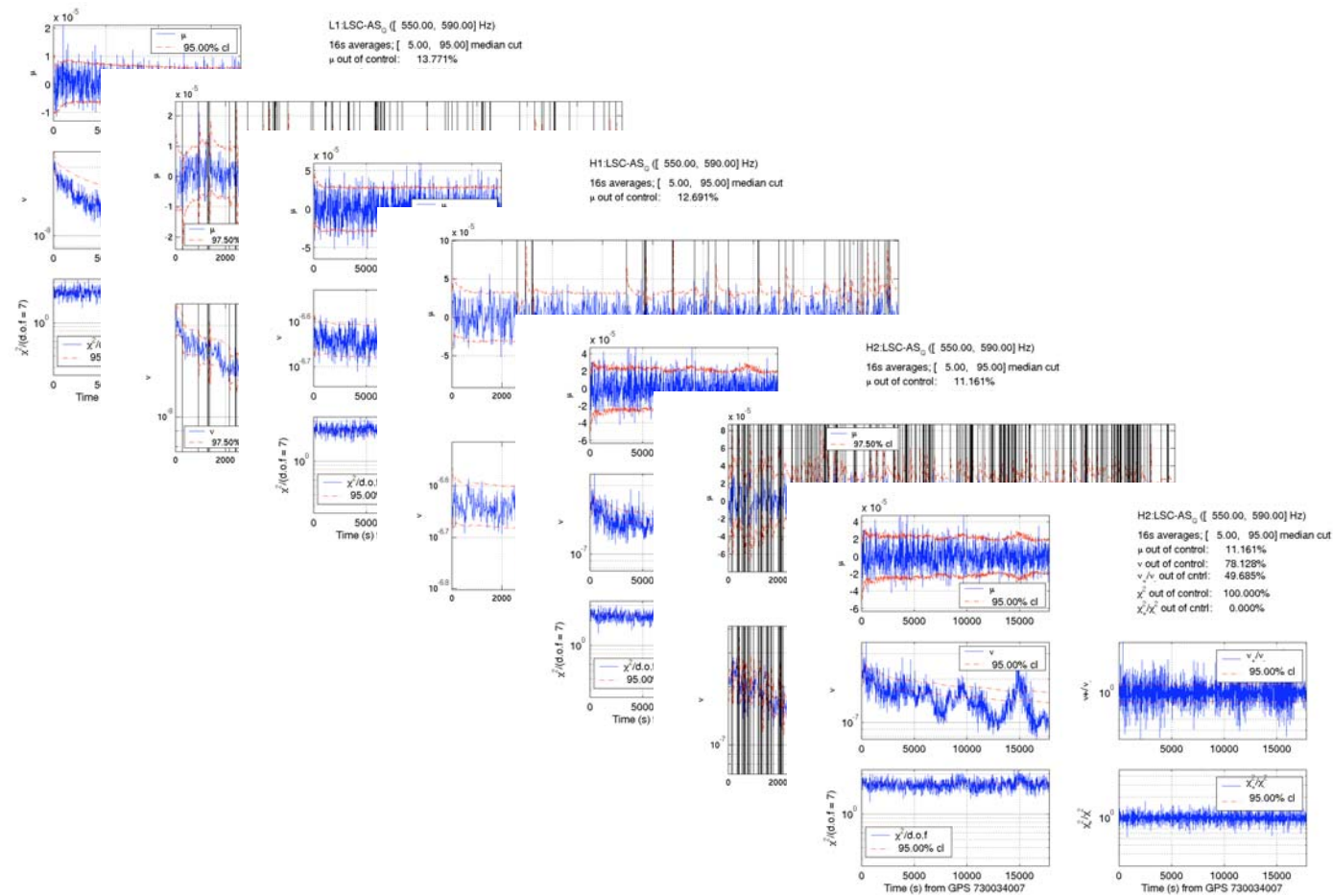
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# Anecdotal look at S2: Longest triple science mode segment



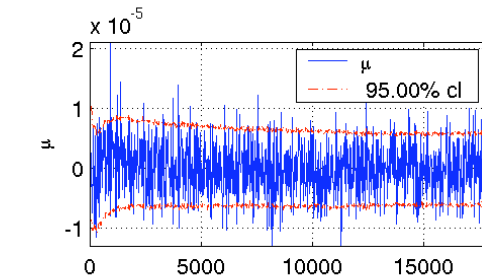
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# L1 Stationarity Monitor: Longest Triple Science Mode



L1:LSC-AS<sub>Q</sub> ([ 550.00, 590.00] Hz)

16s averages; [ 5.00, 95.00] median cut

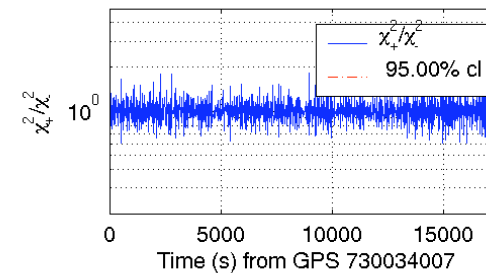
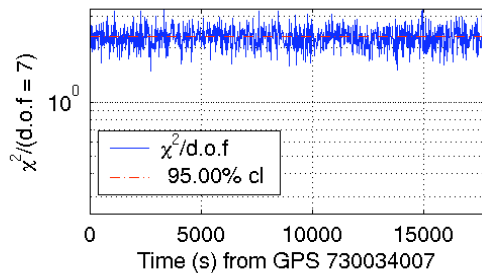
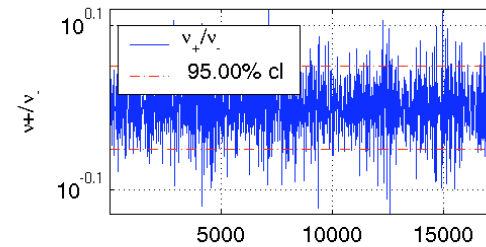
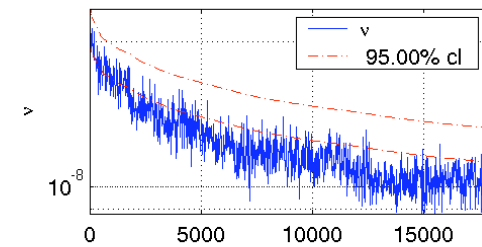
$\mu$  out of control: 13.771%

$v$  out of control: 87.039%

$v_+/v_-$  out of cntrl: 19.082%

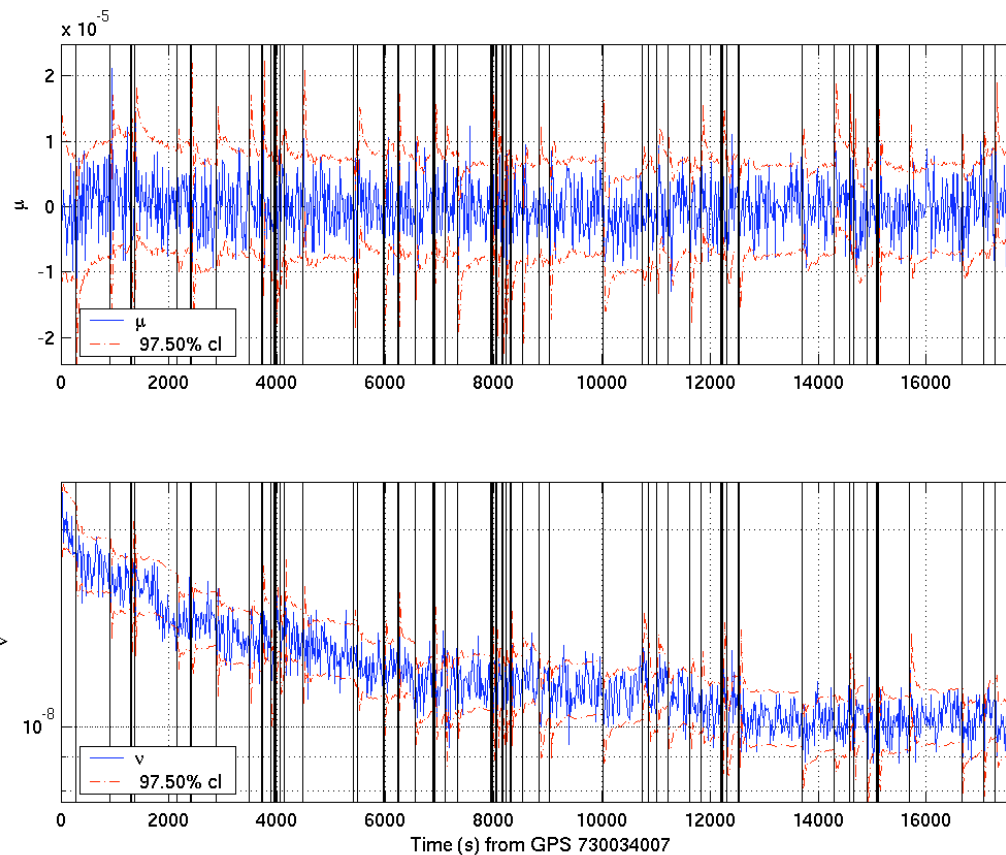
$\chi^2$  out of control: 100.000%

$\chi_+^2/\chi_-^2$  out of cntrl: 0.000%





# L1 Cuts: Longest Triple Science Mode



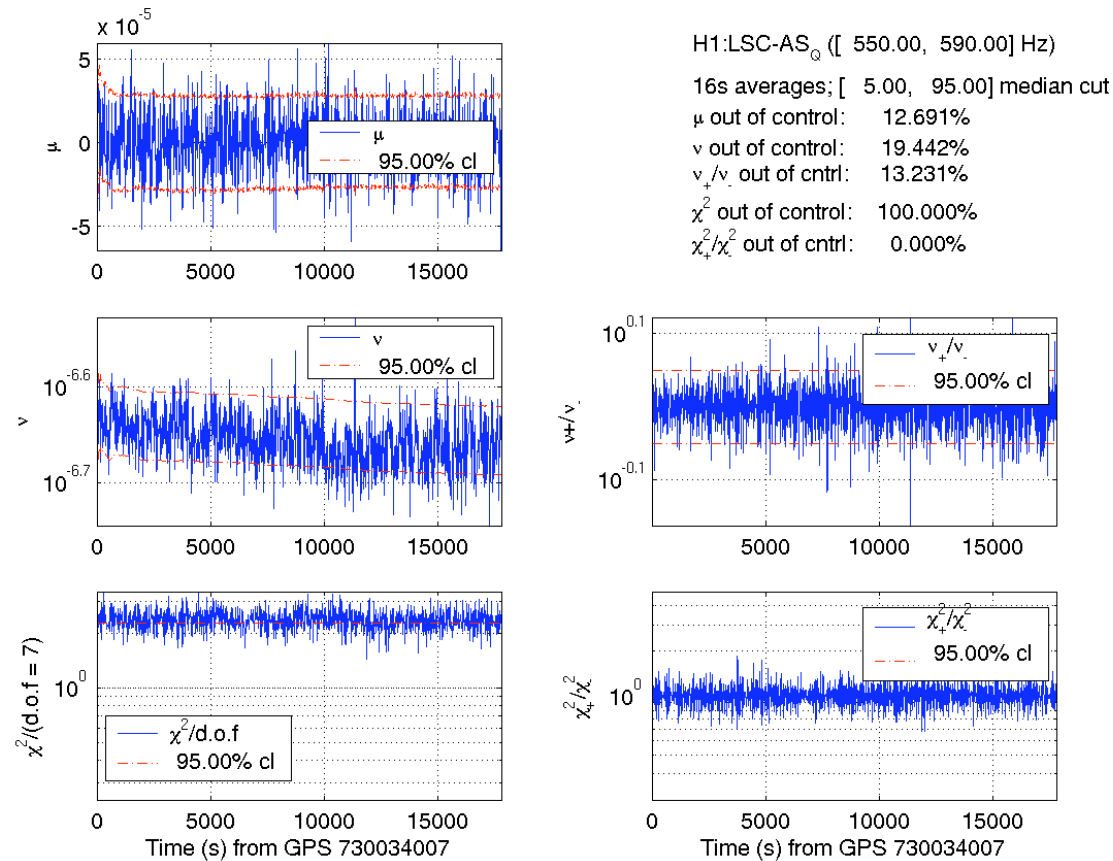
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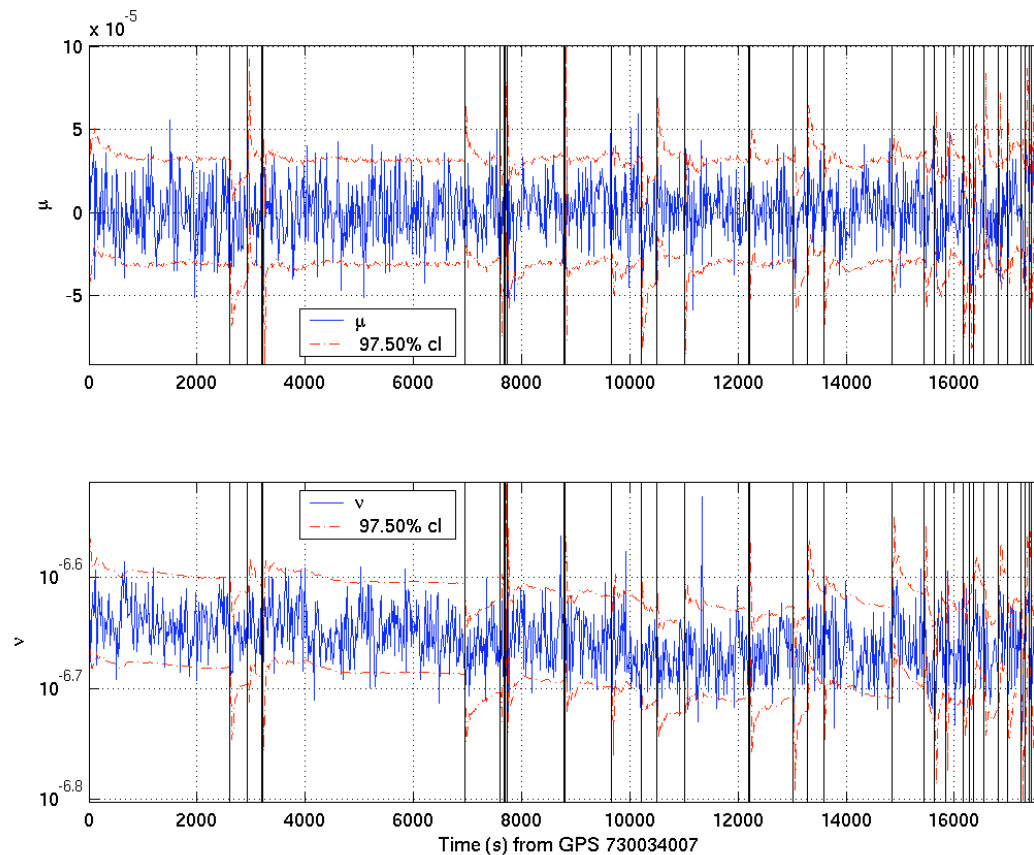


# H1 Stationarity Monitor: Longest Triple Science Mode





# H1 Cuts: Longest Triple Science Mode



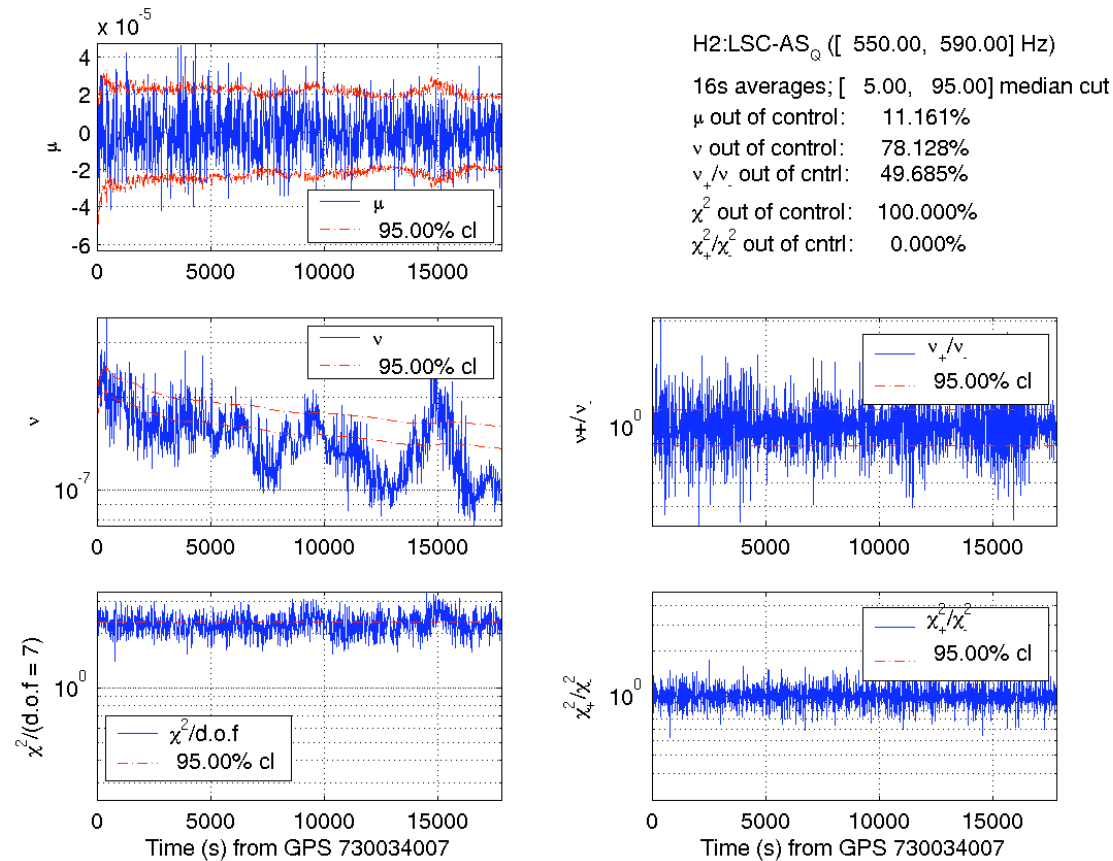
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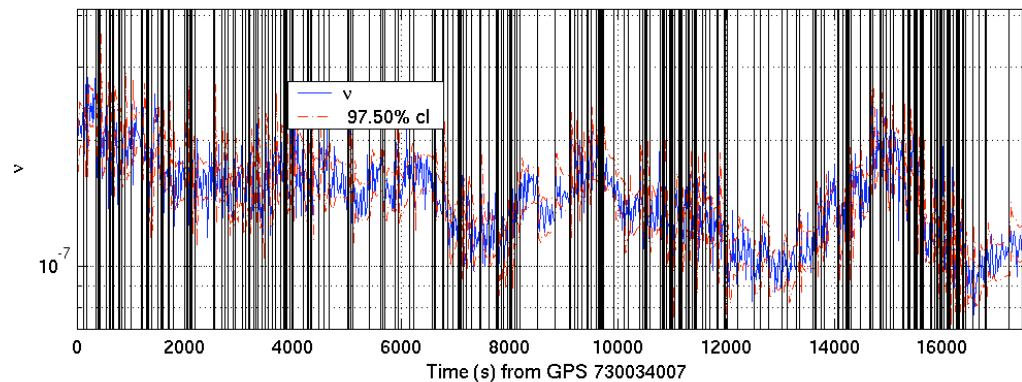
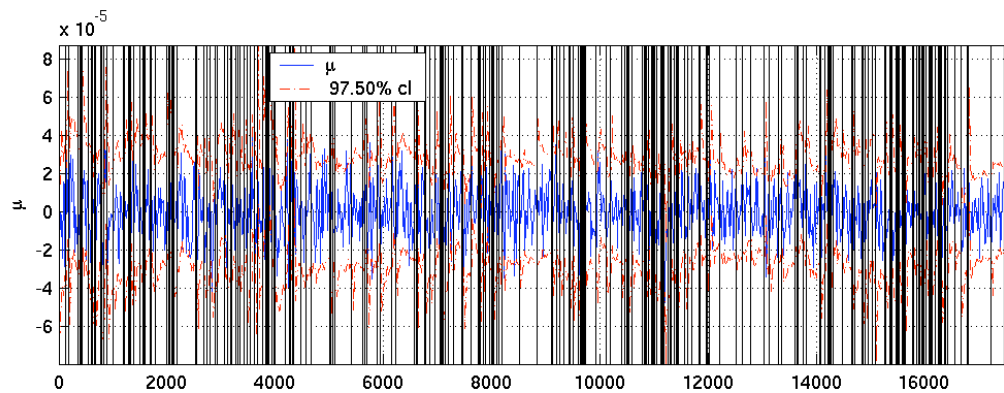


# H2 Stationarity Monitor: Longest Triple Science Mode





# H2 Cuts: Longest Triple Science Mode



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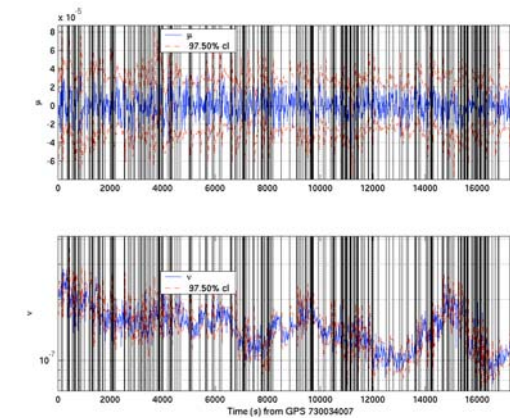
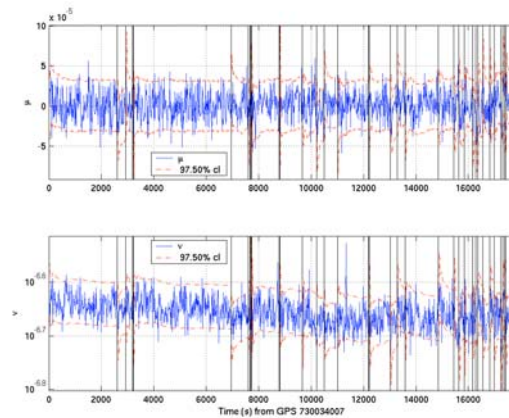
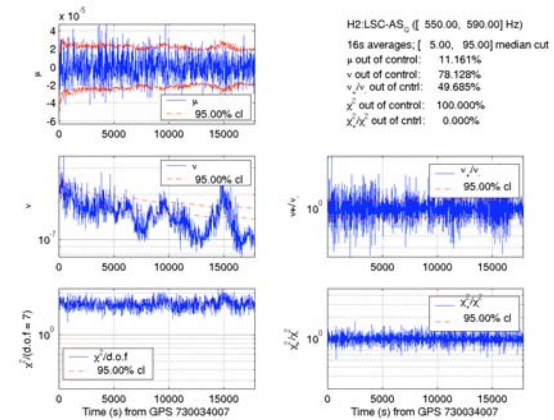
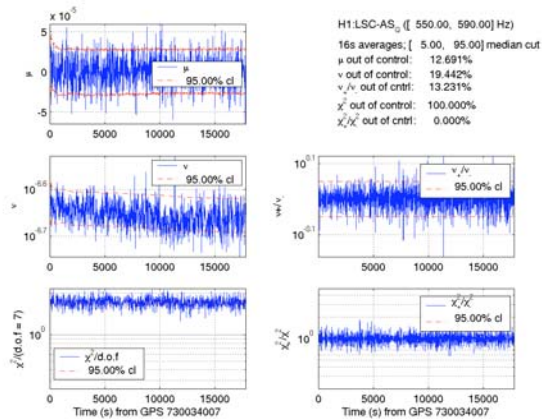
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# H1 & H2: Contrast



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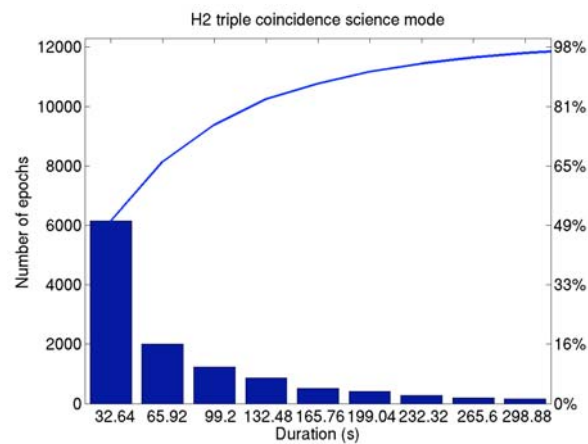
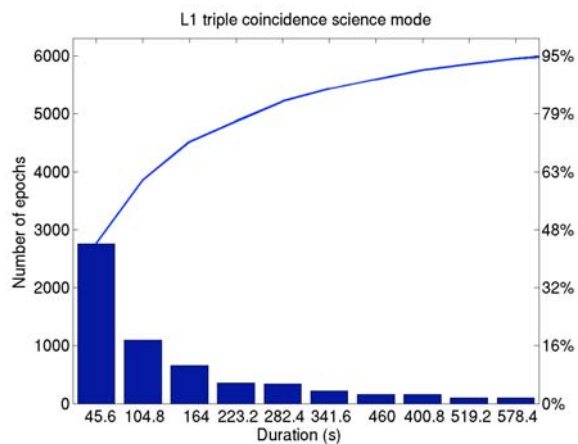
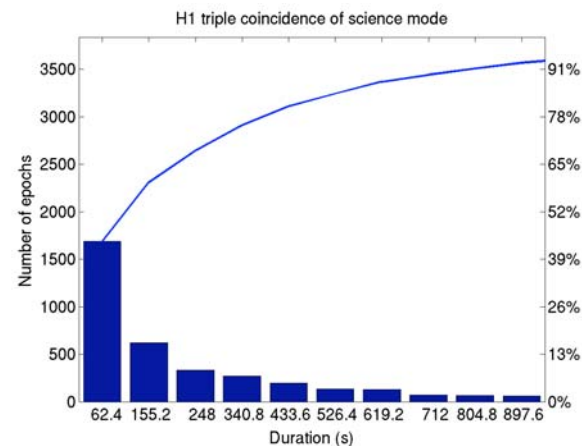
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# Stationary Epoch Durations

- [550 Hz, 590 Hz]
- 16 s averaging time
- [5, 95] percentile cut
- 97.5% confidence stationarity criteria



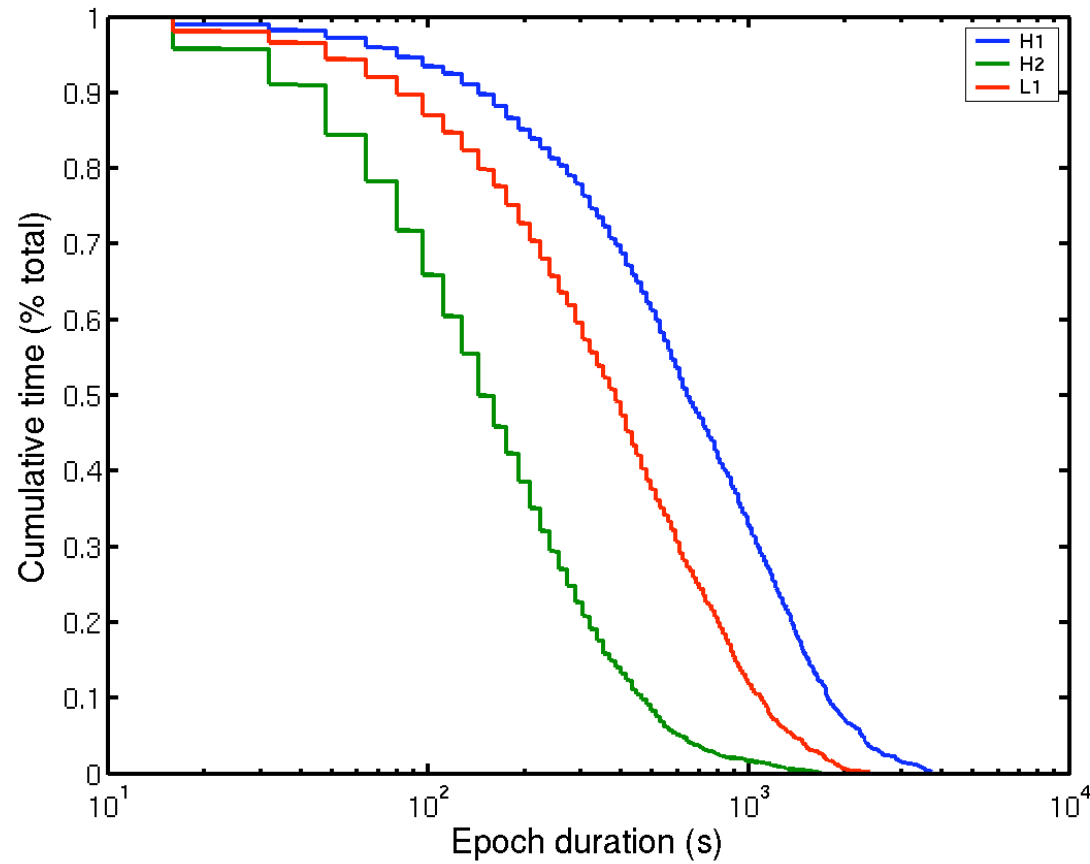
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# Stationary Epoch Durations





# Summary

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- S2 noise shows **strongly variable, frequent outliers**
  - » Trimming samples outside of [5,95] percentile on 16s data in [550,590] Hz band reduces range of variance by factor 35 or more
- S2 noise shows **strong variability on minute** timescales
  - » 50% of H2/L1/H1 *epochs* shorter than 64/96/144 s: expect median durations of ~750s for Gaussian stationary noise
  - » 50% of H2/L1/H1 *durations* in epochs shorter than 144/384/640 s: expect ~1780 s for Gaussian stationary noise
- Science results seriously affected by noise non-stationarity
  - » Sophisticated analyses can account for variability *if not too rapid*
  - » Greater, more rapid variability, weaker conclusions
- medNoiseChar available as Matlab script and as a standalone Ap that runs on frame data
  - » Same concepts being applied to calibration lines: cf. lineamp being developed by Mike Ashley