

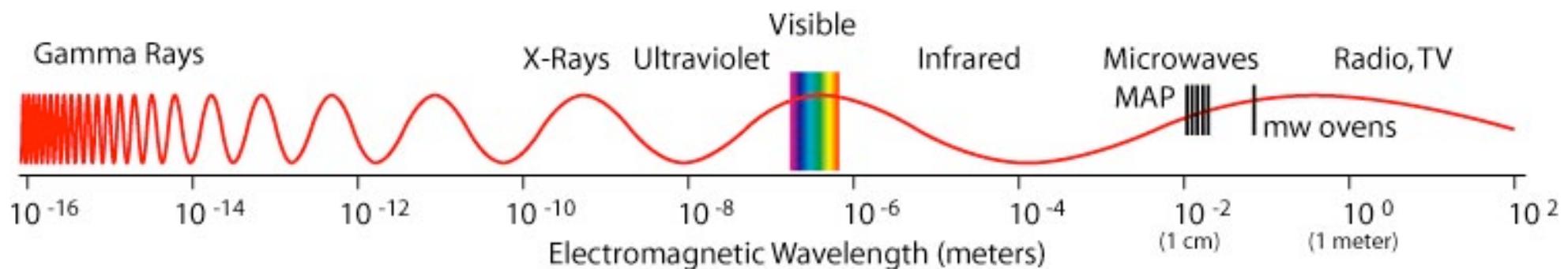
Gravitational Wave Astronomy: Needle in a Haystack

Neil Cornish
Montana State University

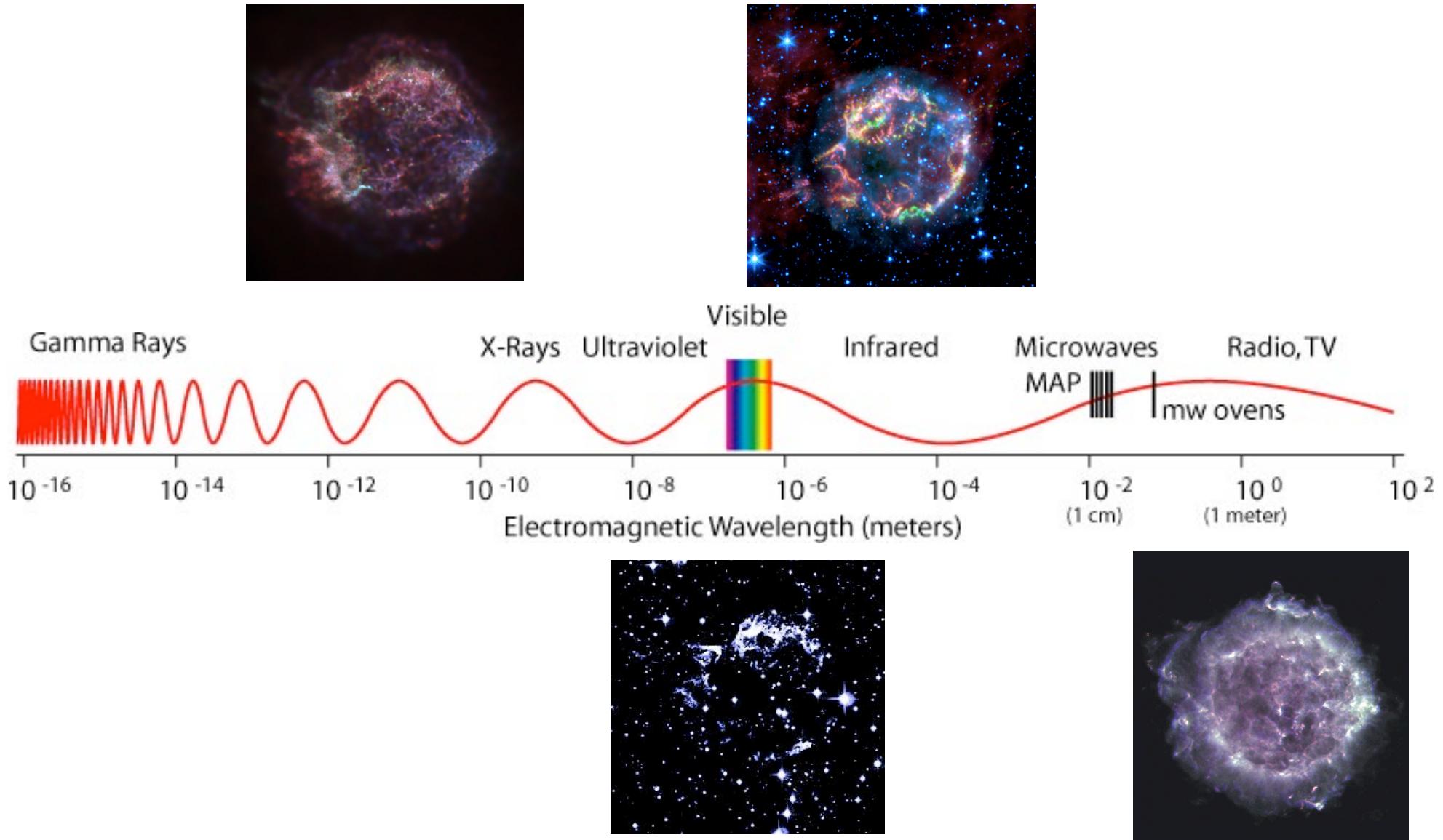




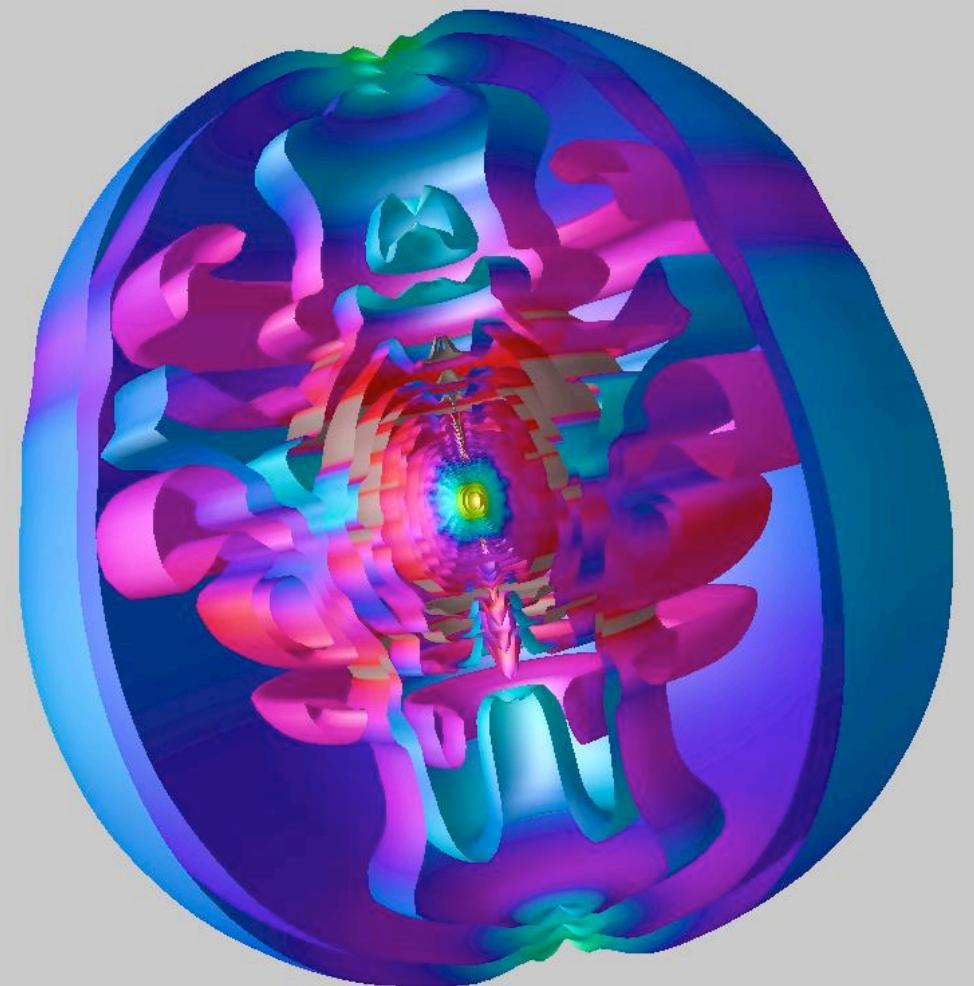
Astronomy Last Century: Multi-Wavelength



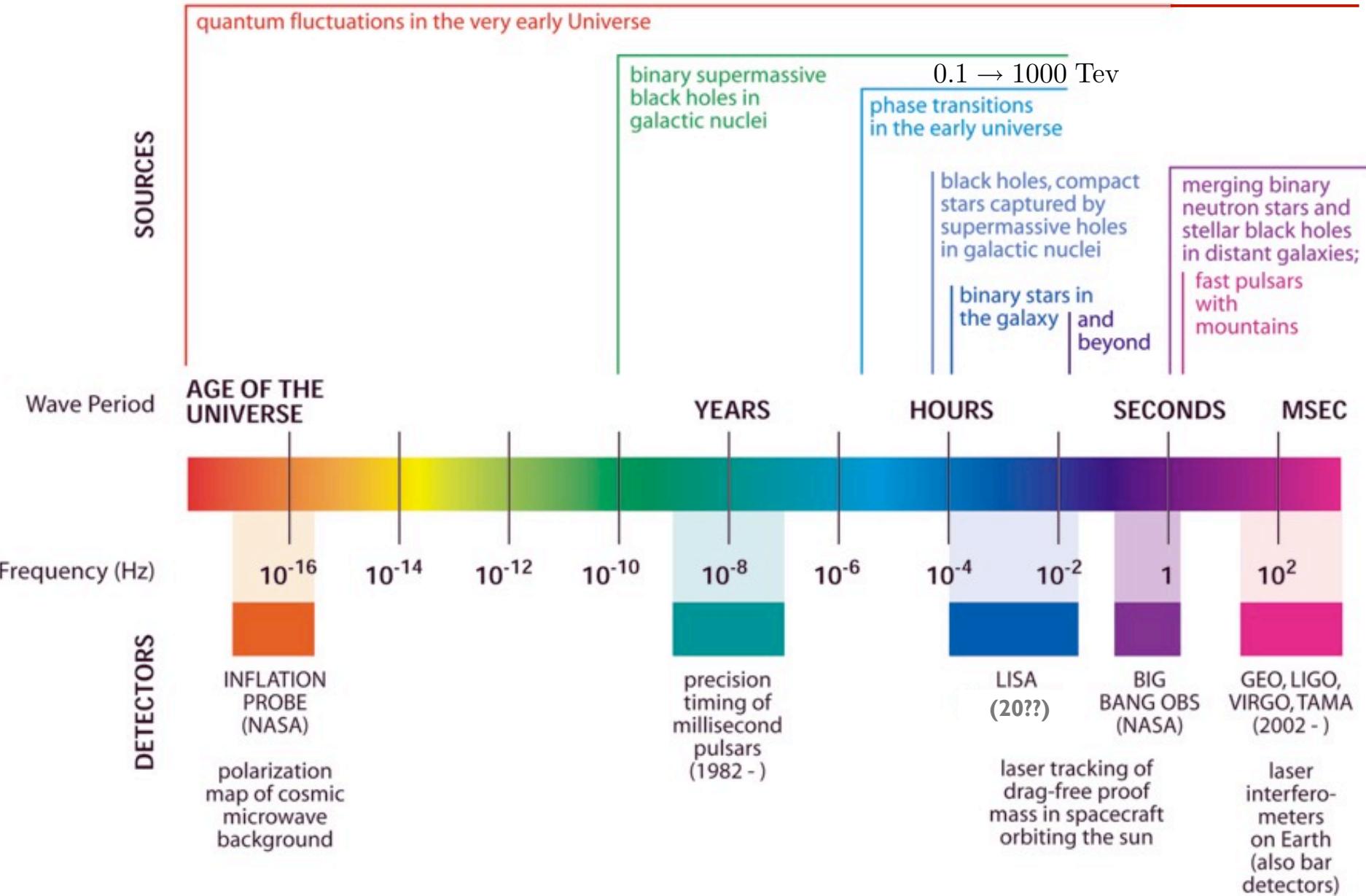
Astronomy Last Century: Multi-Wavelength

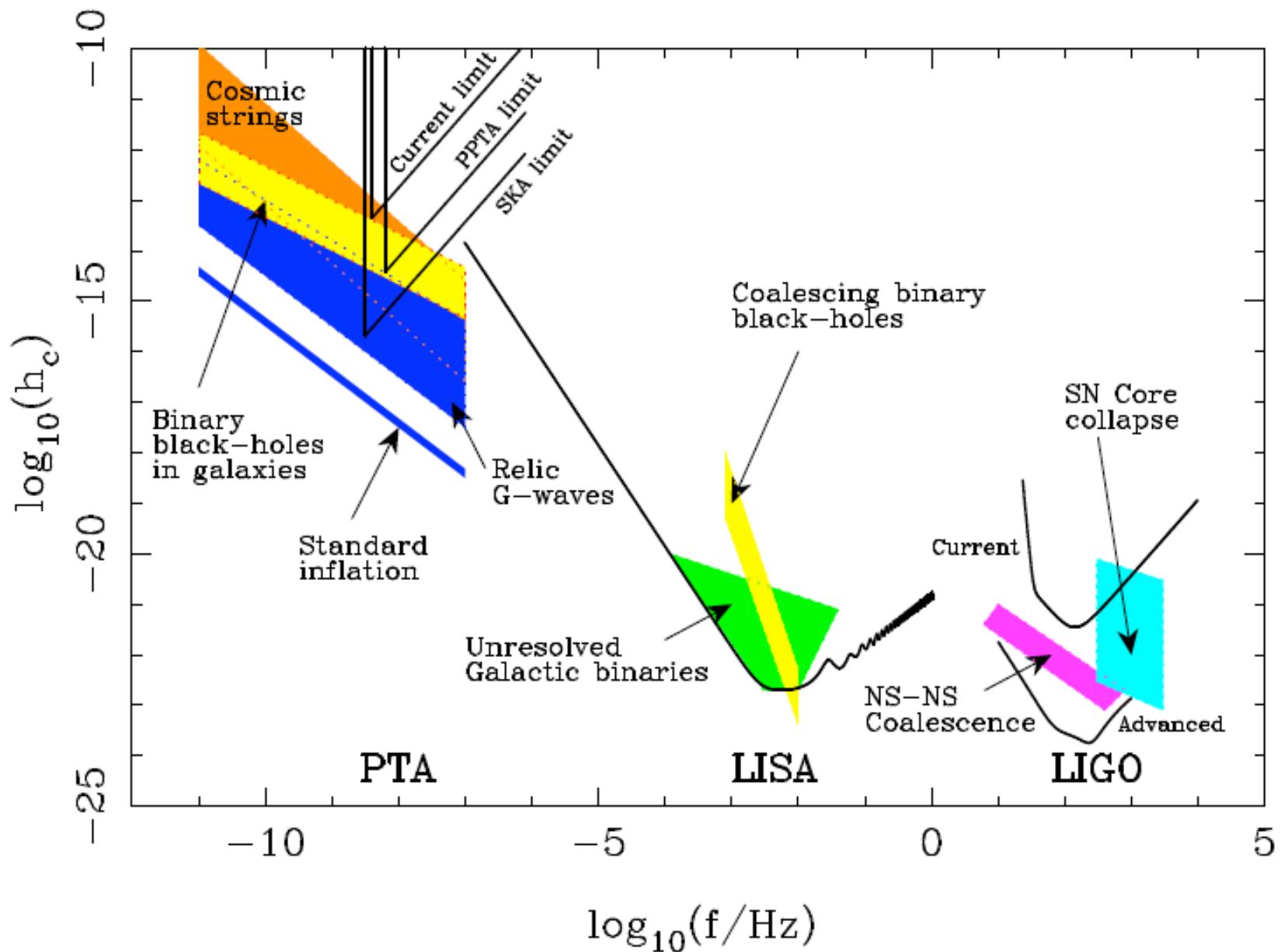


Astronomy this Century: Multi-Spectrum



THE GRAVITATIONAL WAVE SPECTRUM





Recipe for Making Gravitational Waves

Ingredients:

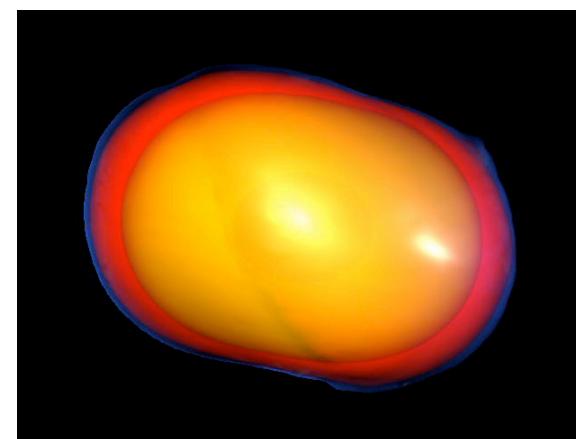
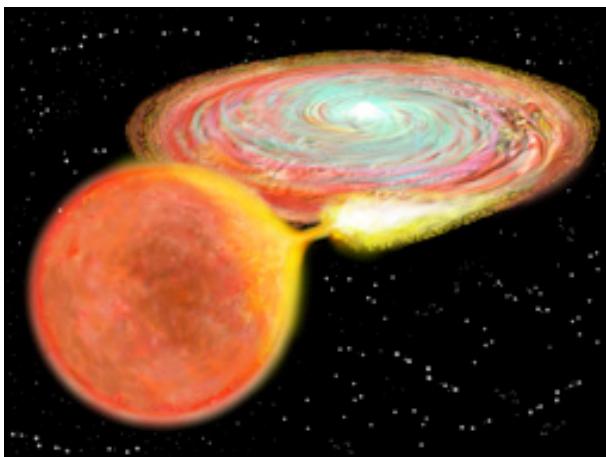
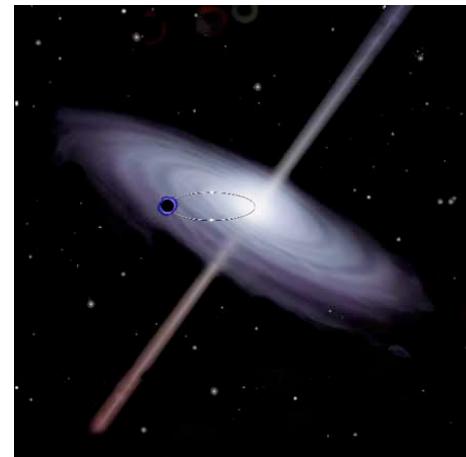
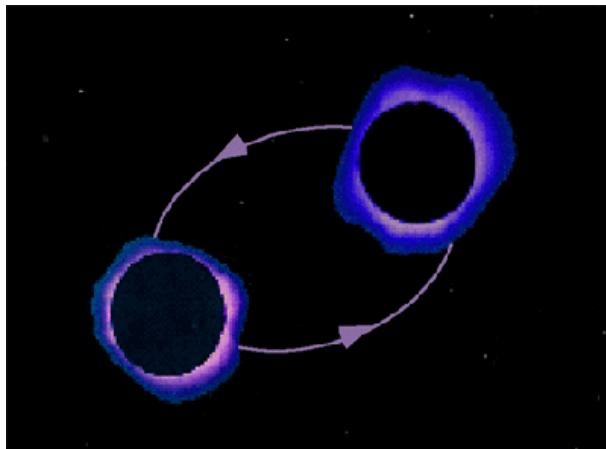
- Large amount of mass/energy
(any type will do).

Directions:

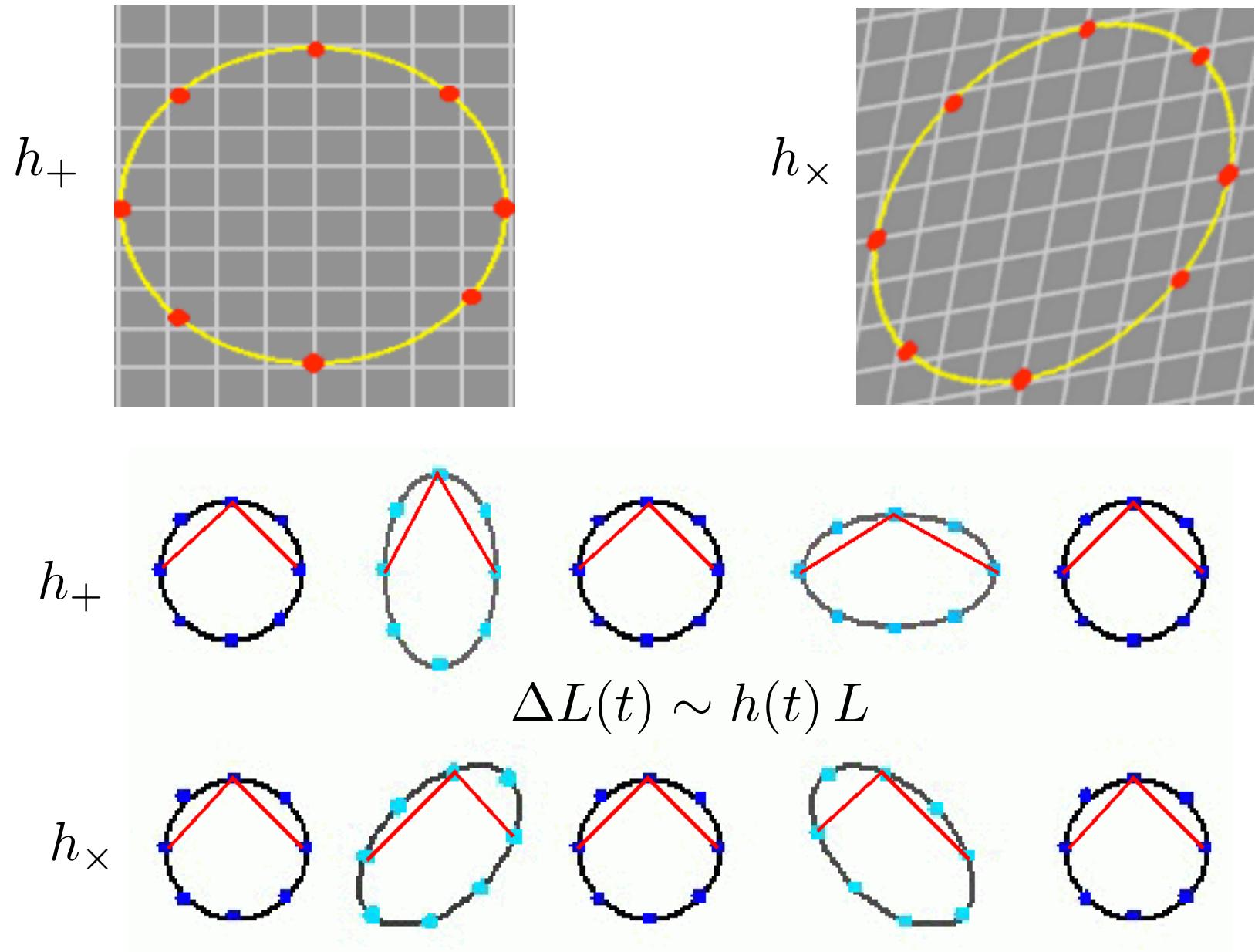
- Squish into a small, lumpy blob.
(two blobs work better)
- Shake or stir vigorously.



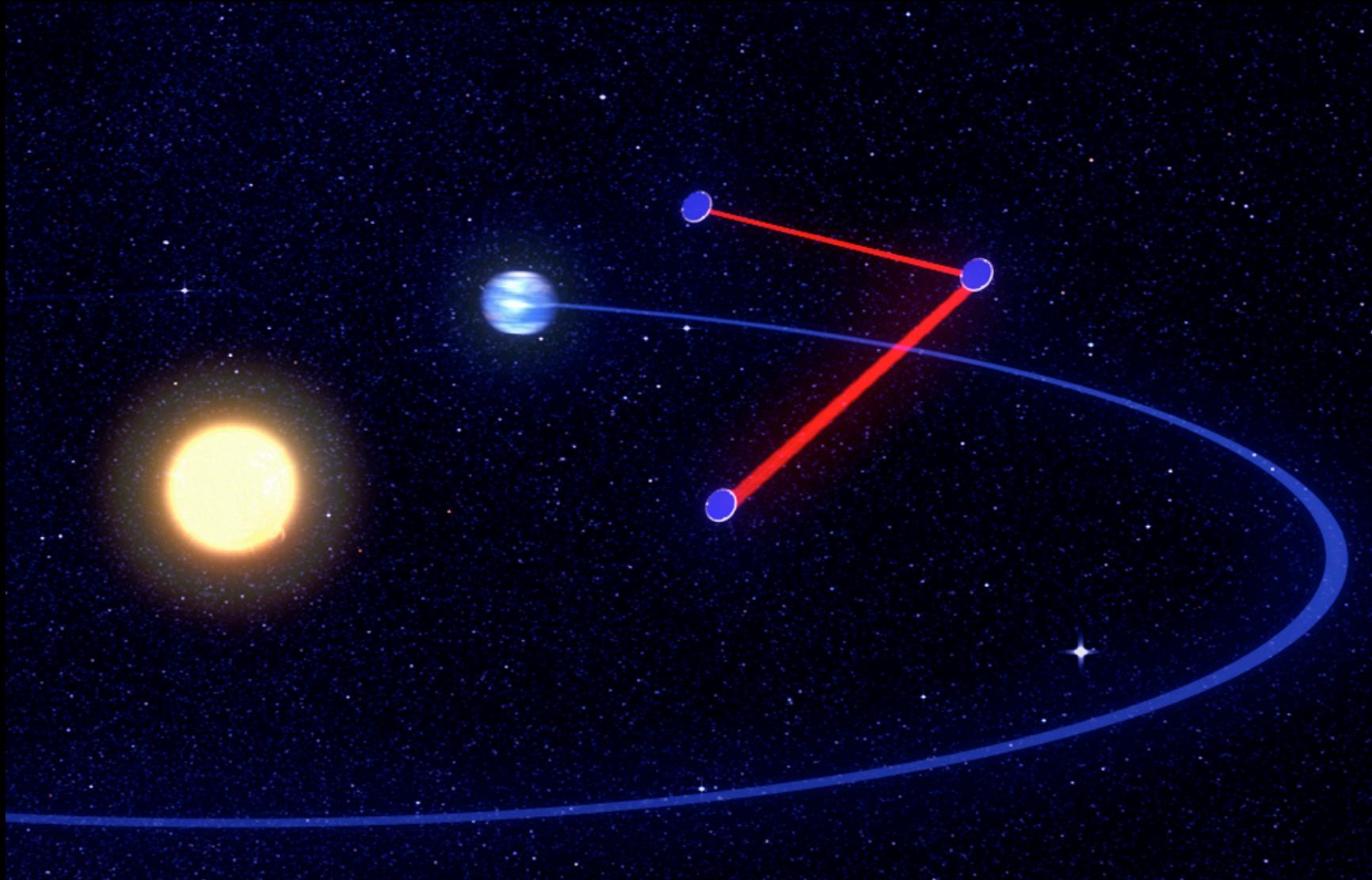
Gravitational Wave Sources



Detecting Gravitational Waves



Laser Interferometer Space Antenna



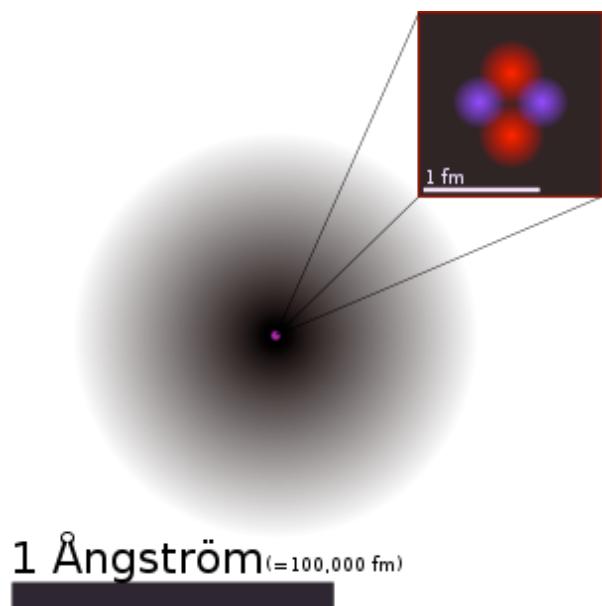
eLISA - <http://elisa-ngo.org/>

Laser Interferometer Gravitational Observatory

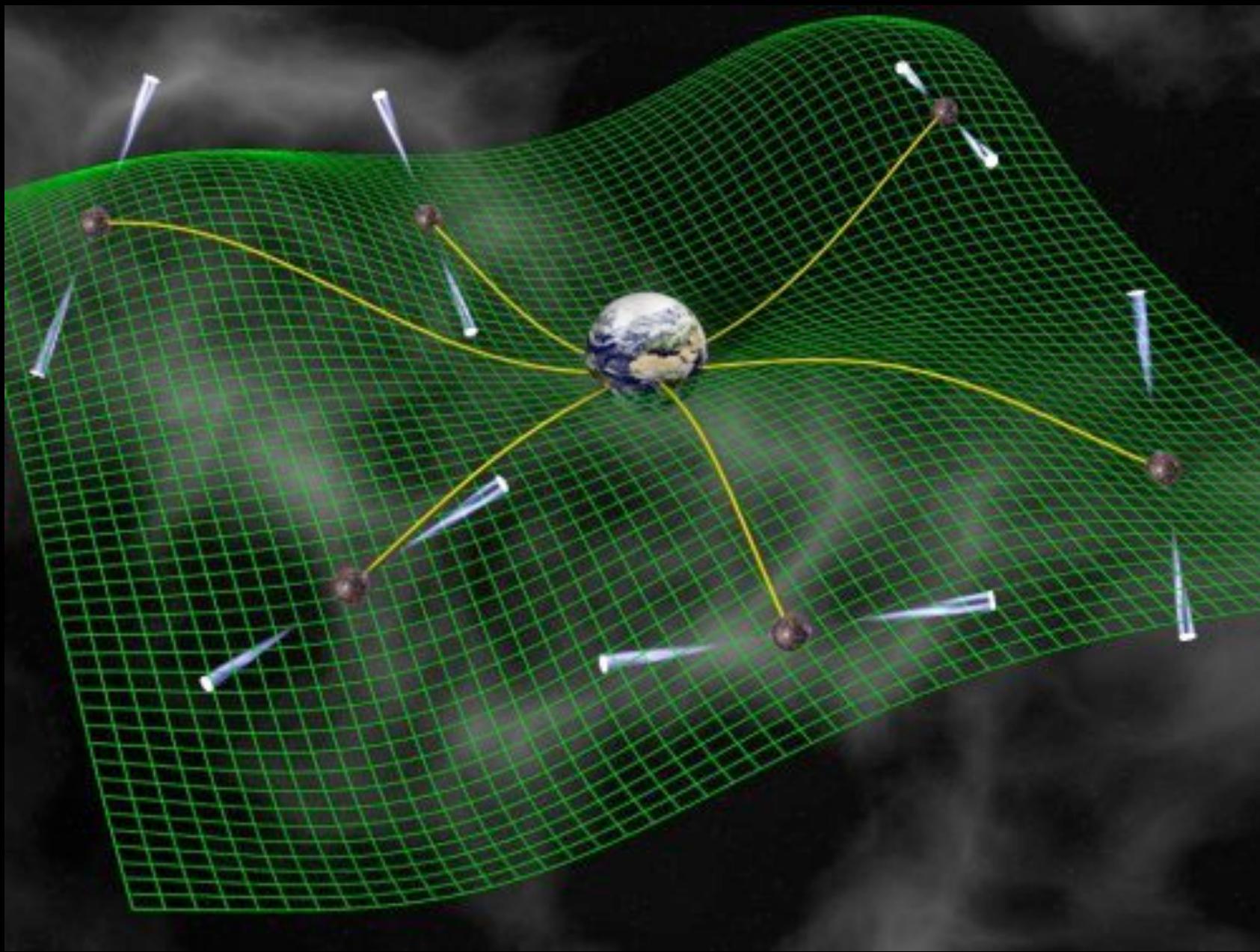


$$h = \frac{\Delta L}{L} \sim 10^{-21}$$

$$L = 4 \text{ km} \Rightarrow \Delta L = 10^{-3} \text{ fm}$$



Pulsar Timing Array



Signal Analysis

$$s_i(t) = R_i(t, \tau) \star h(\tau) + n_i(t)$$

Detector Output

Detector Response

GW Strain

Detector Noise

e.g. single LIGO detector

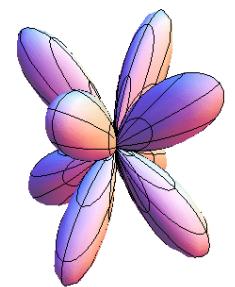
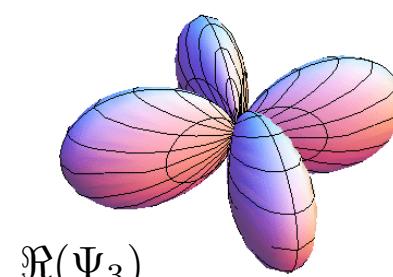
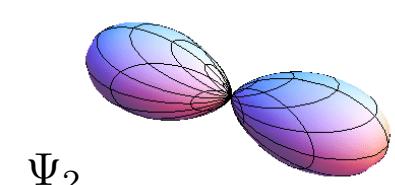
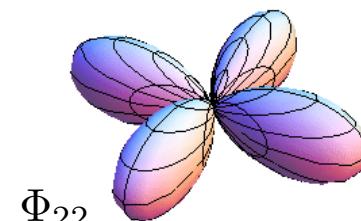
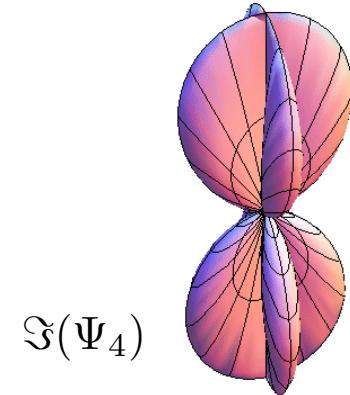
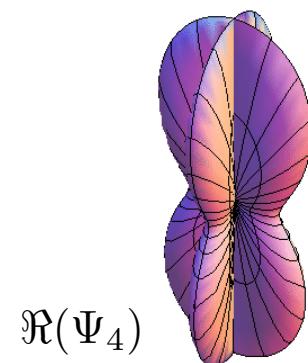
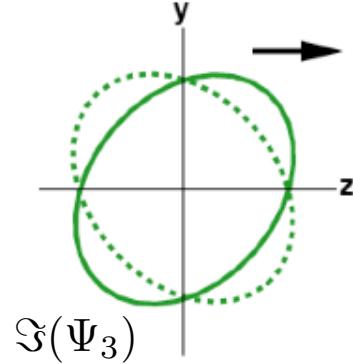
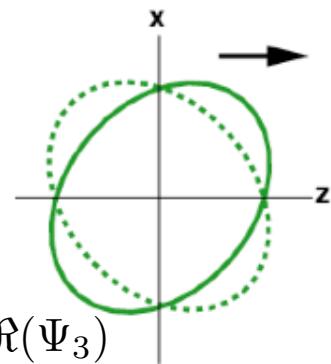
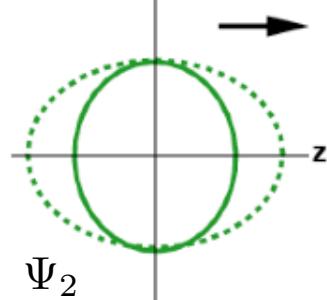
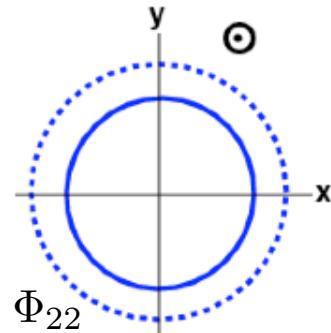
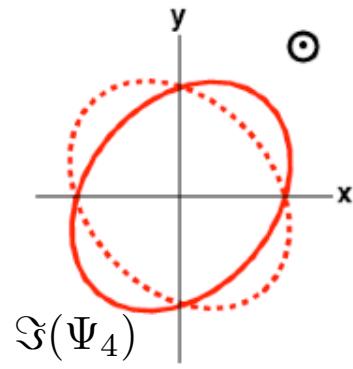
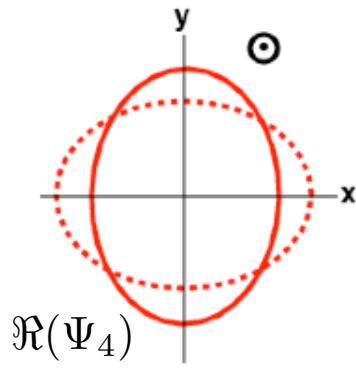
$$s(t) = F^+(t)h_+(t - \tau) + F^\times(t)h_\times(t - \tau) + n(t)$$

More abstractly:

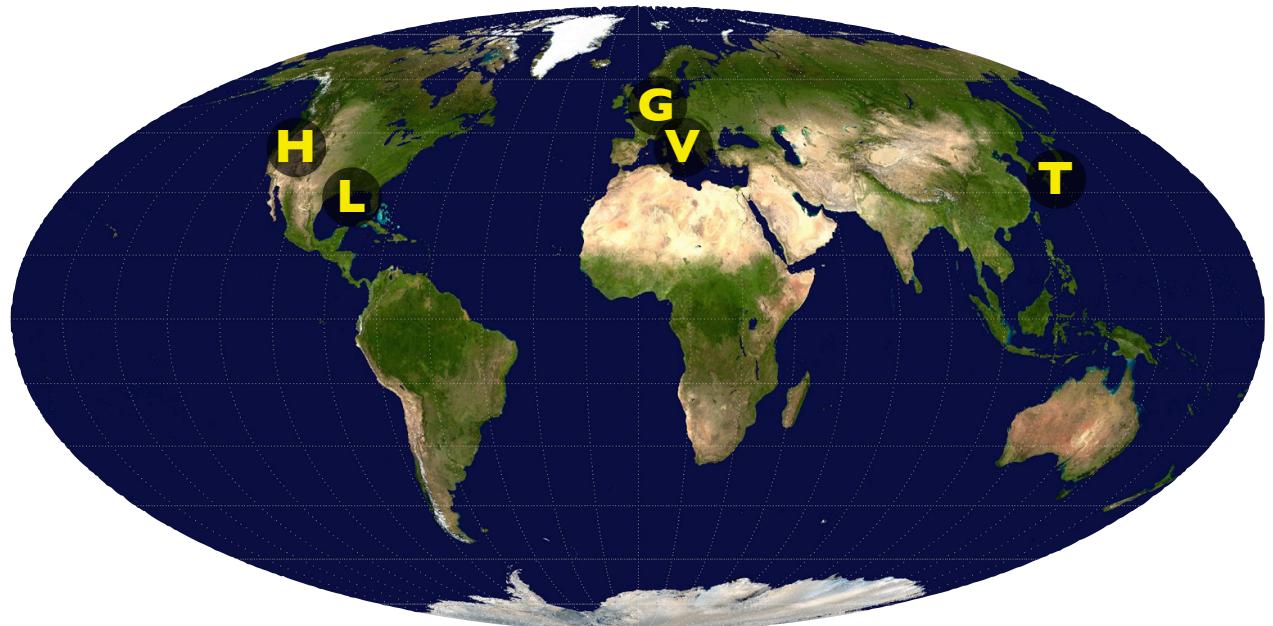
$$\mathbf{s} = \mathbf{R} \cdot \mathbf{h} + \mathbf{n}$$

Alternative Theories Predict Additional Polarization States

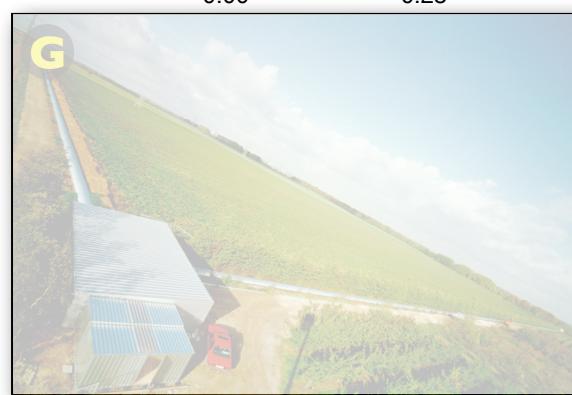
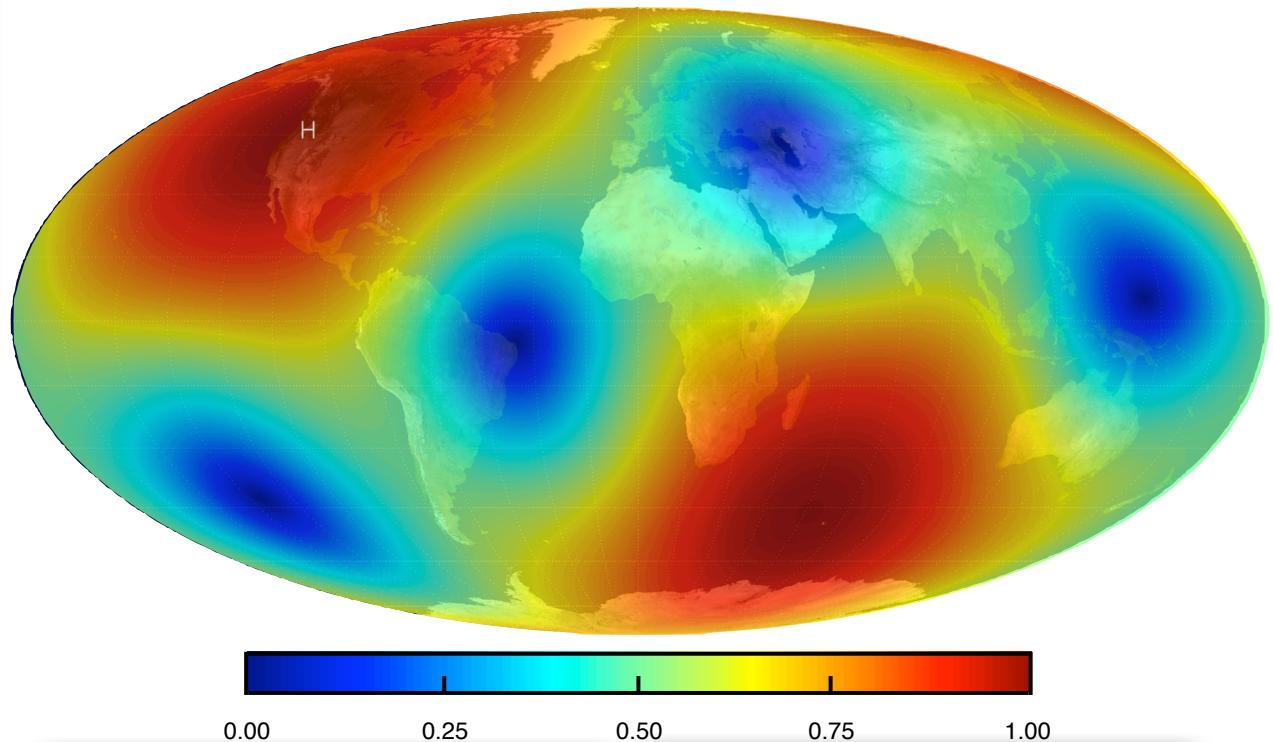
Gravitational-Wave Polarization



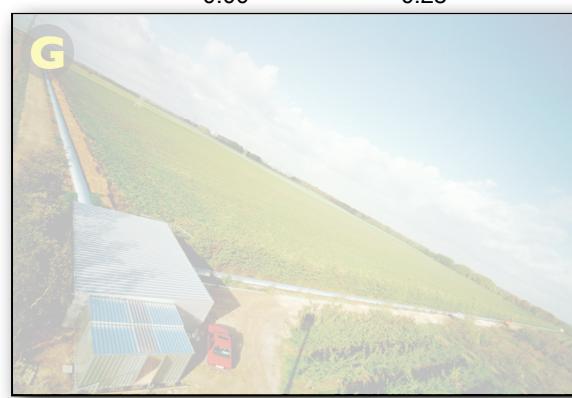
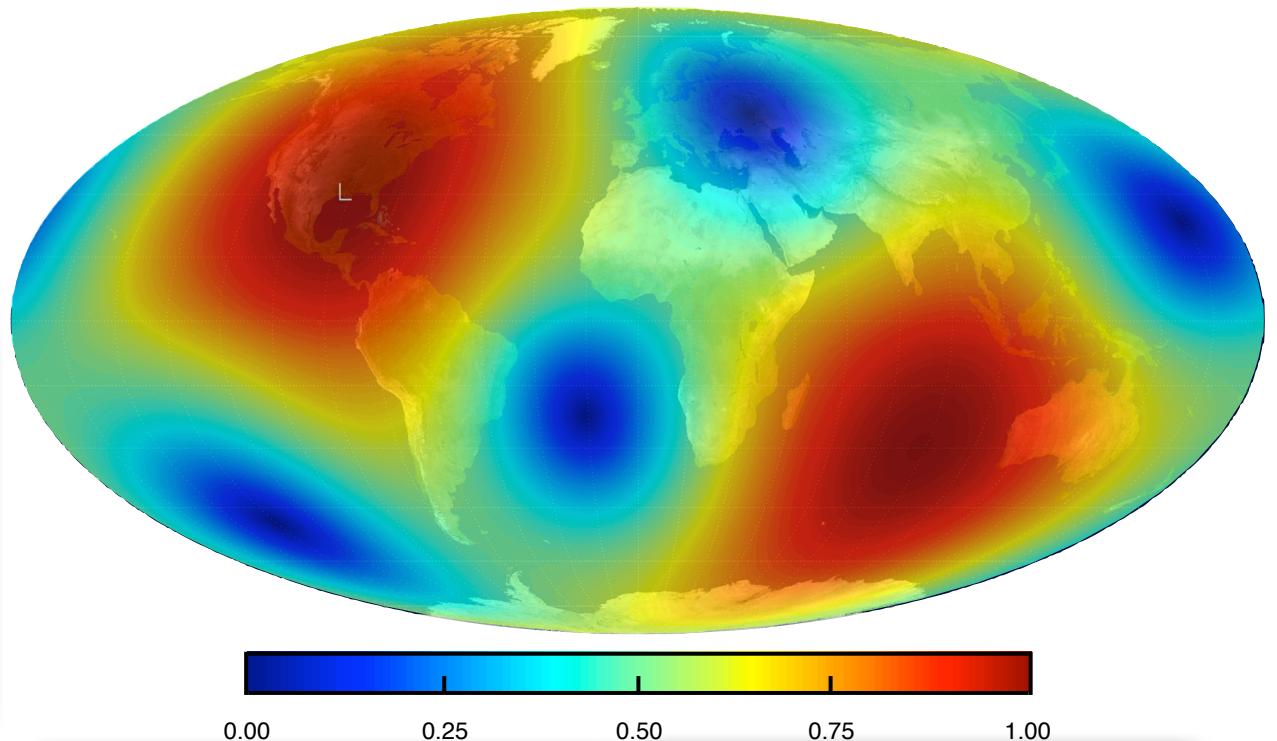
Gravitational Wave Detectors



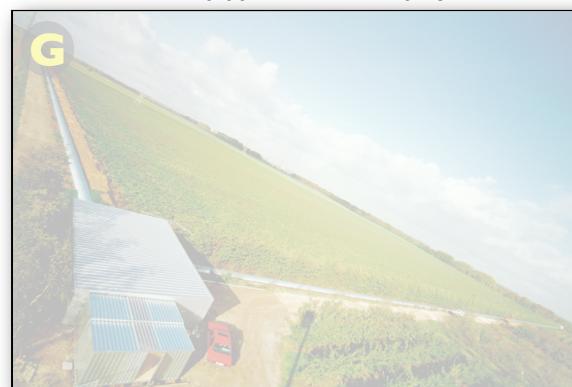
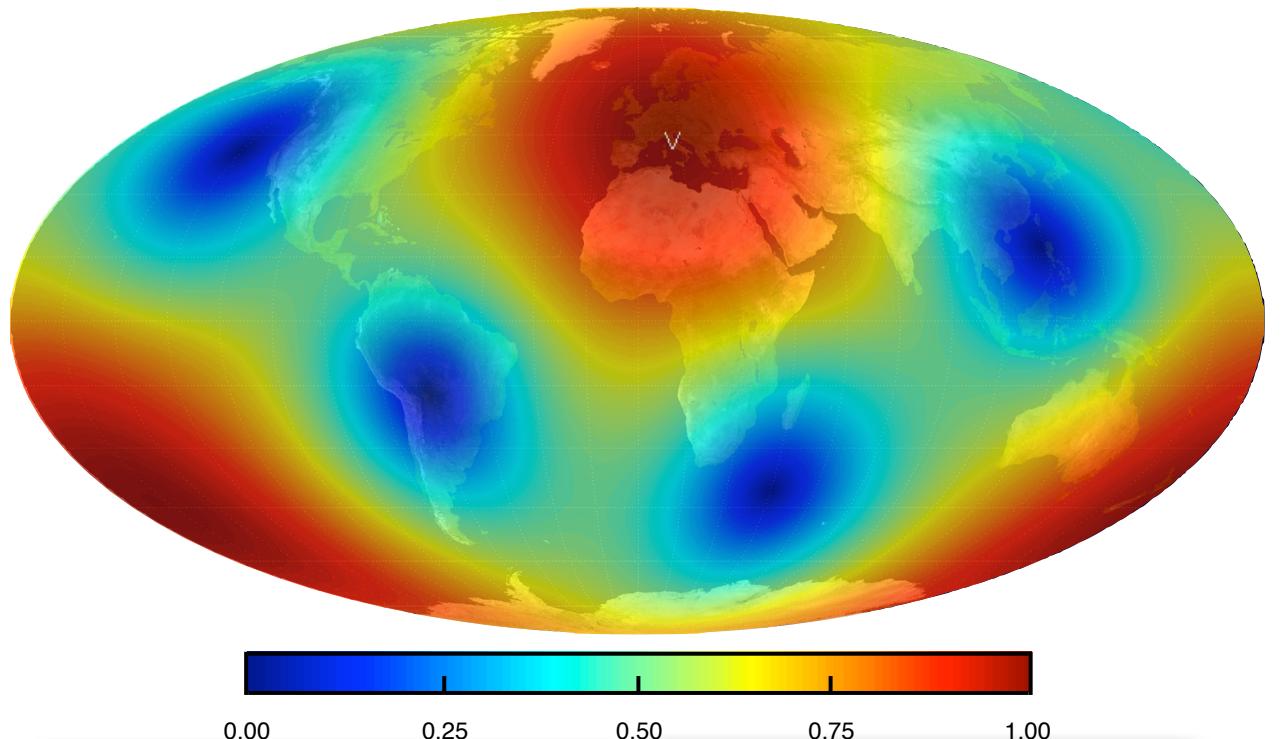
Gravitational Wave Detectors



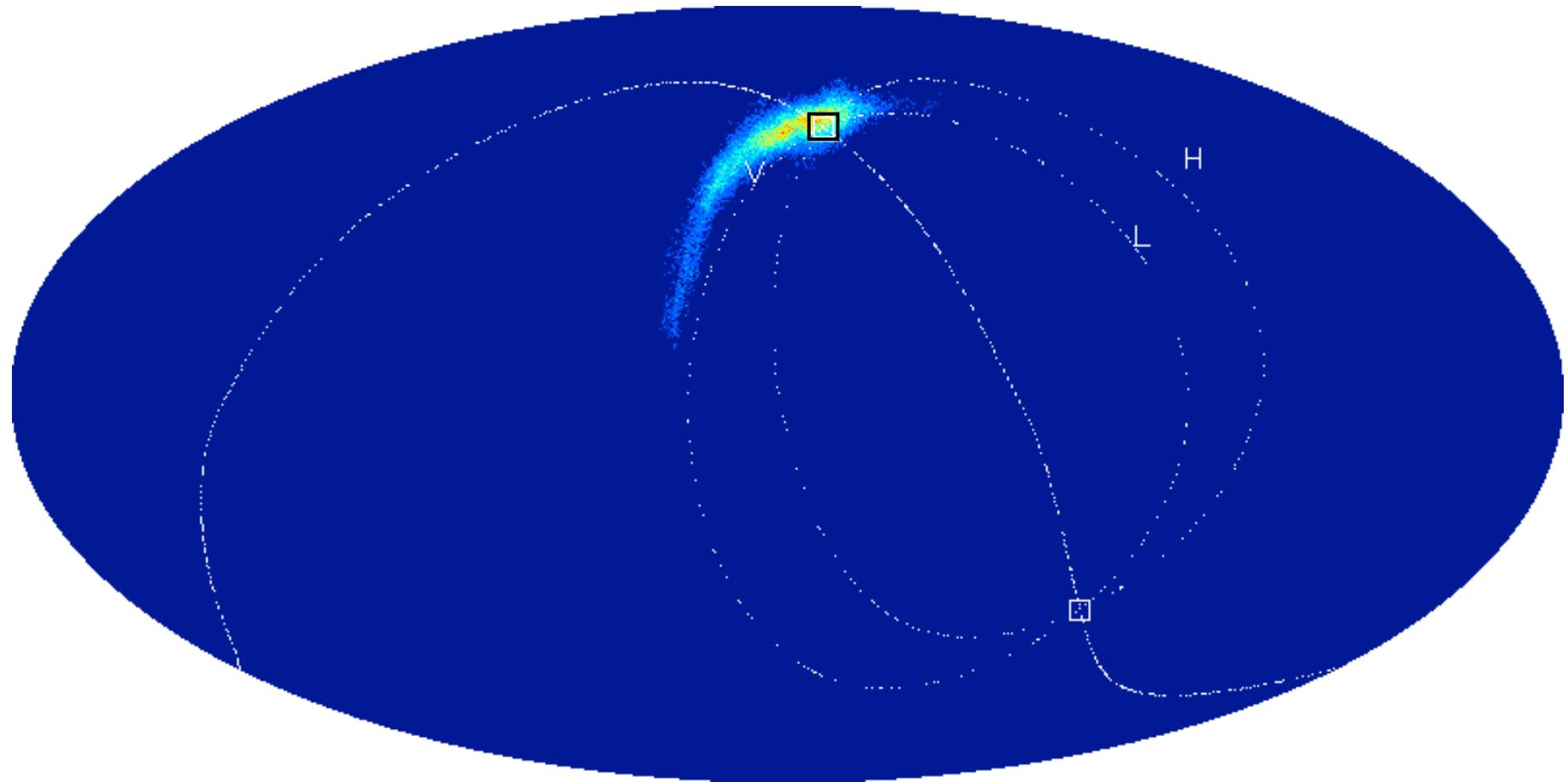
Gravitational Wave Detectors



Gravitational Wave Detectors



Triangulation



Signal Analysis

$$\mathbf{s} = \mathbf{R} \cdot \mathbf{h} + \mathbf{n}$$

Main analysis techniques:

- (1) Direct Model Inference
- (2) Channel Cross Correlation
- (3) Inference using Summary Statistics

Model Inference

$$\mathbf{s} = \mathbf{R} \cdot \mathbf{h}_A + \mathbf{n}$$

Form residual: $\mathbf{r} = \mathbf{s} - \mathbf{R} \cdot \mathbf{h}$

Demand that residual is consistent with instrument noise:

\Rightarrow Likelihood $p(\mathbf{s}|\mathbf{h}) = p_n(\mathbf{s} - \mathbf{R} \cdot \mathbf{h})$

Classical Statistics

Neyman Pearson optimal statistic

$$\Lambda = \frac{p(\mathbf{s}|\mathbf{h})}{p(\mathbf{s}|0)}$$

Maximized w.r.t amplitude yields SNR ratio statistic

$$\Rightarrow \rho = \frac{(\mathbf{s}|\mathbf{R} \cdot \mathbf{h})}{(\mathbf{R} \cdot \mathbf{h}|\mathbf{R} \cdot \mathbf{h})^{1/2}}$$

“Weiner matched filtering” - optimal statistic for known signal in stationary, Gaussian noise. Neither of which pertain.

Model Inference

$$\mathbf{s} = \mathbf{R} \cdot \mathbf{h}_A + \mathbf{n}$$

Need models for the instrument response, noise and GW signals

Instrument response 

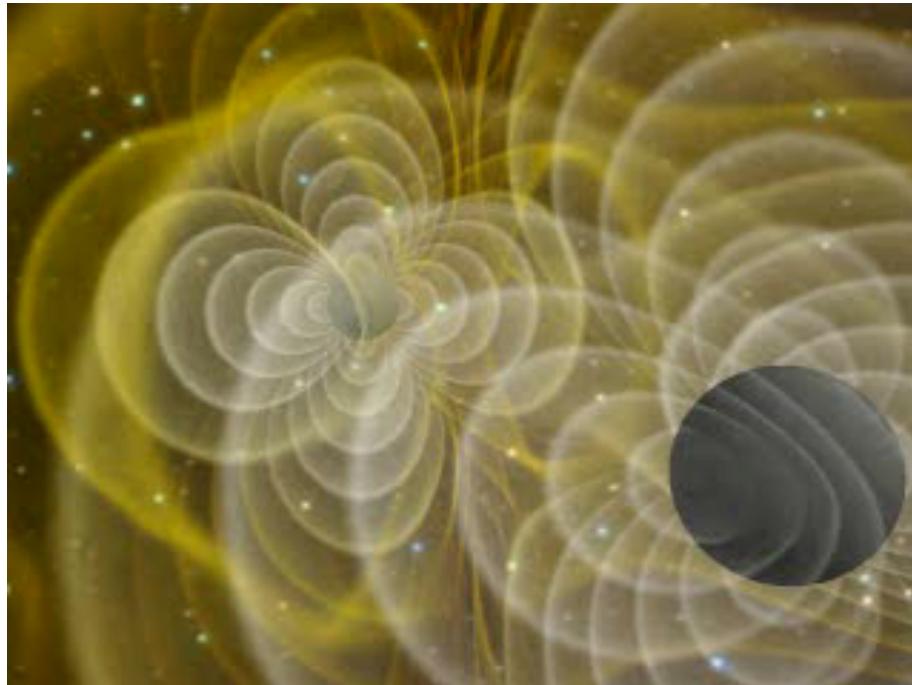
Instrument noise ?

GW signals ? Some better than others

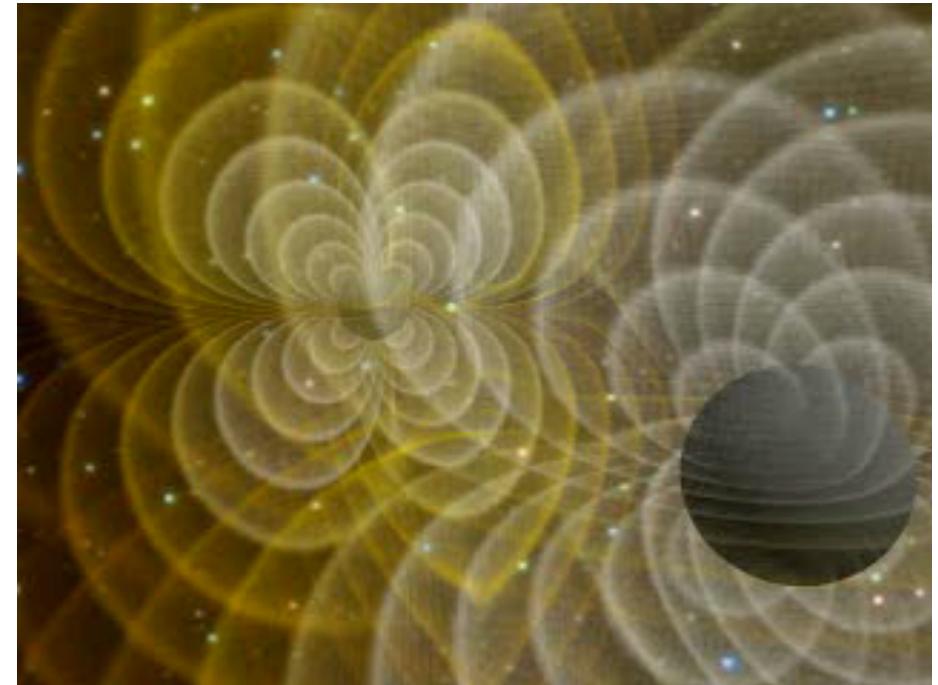
e.g. Black hole binaries $\mathbf{h}(\vec{\lambda})$

$\vec{\lambda}$ 17-dimensional, includes sky location, merger time, masses, spins, orbital eccentricity etc

Modeling BH Mergers



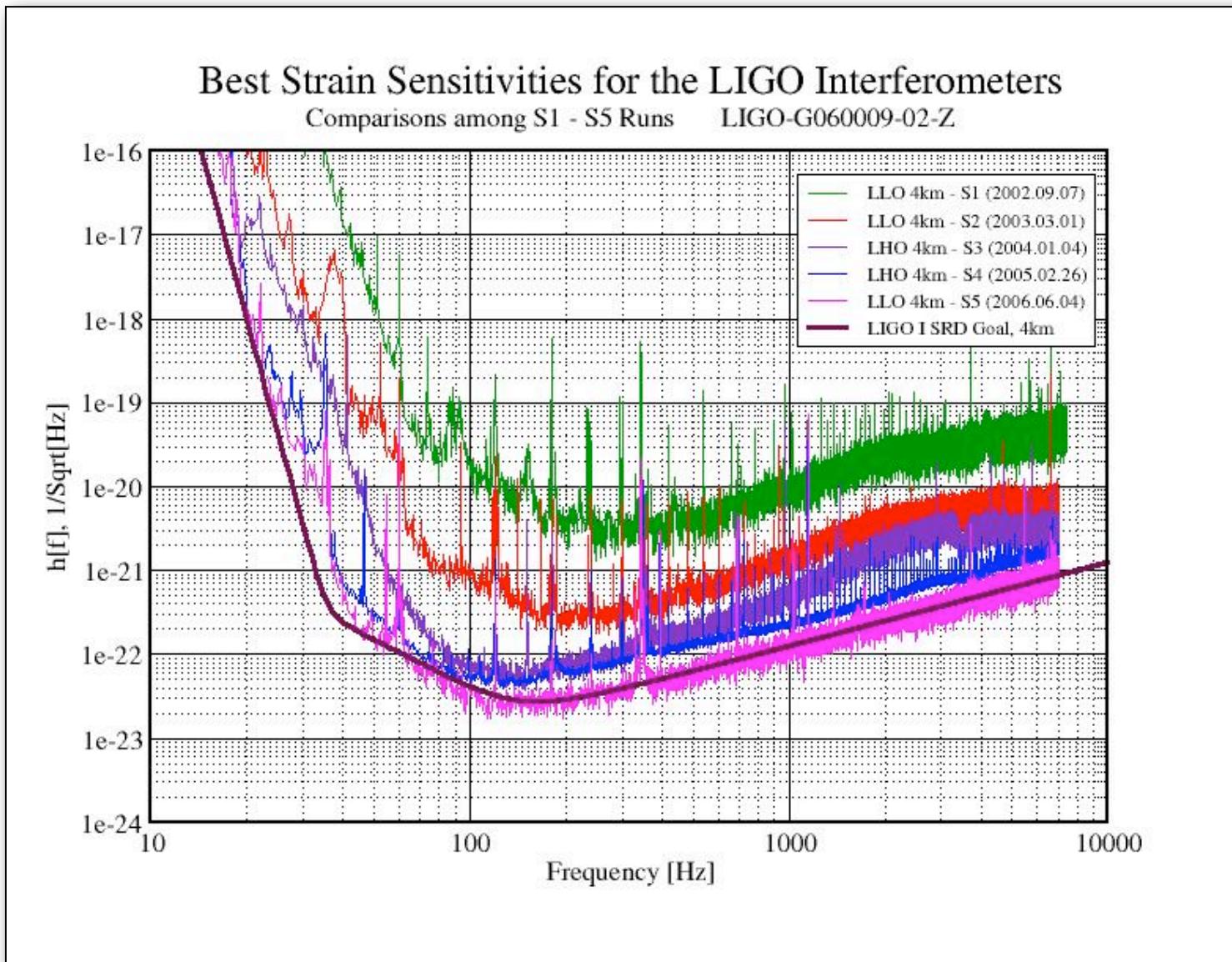
Contours: Curvature
components corresponding to
“+” polarization of GWs



Contours: Curvature
components corresponding to
“x” polarization of GWs

Movies courtesy GSFC Numerical Relativity Group

Example LIGO Noise Spectra



Standard (Naive) Noise Model

Stationary, coloured,
Gaussian noise

$$E[n(f)] = 0$$

$$E[n(f)n^*(f')] = \frac{T}{2}\delta_{ff'}S_n(f)$$

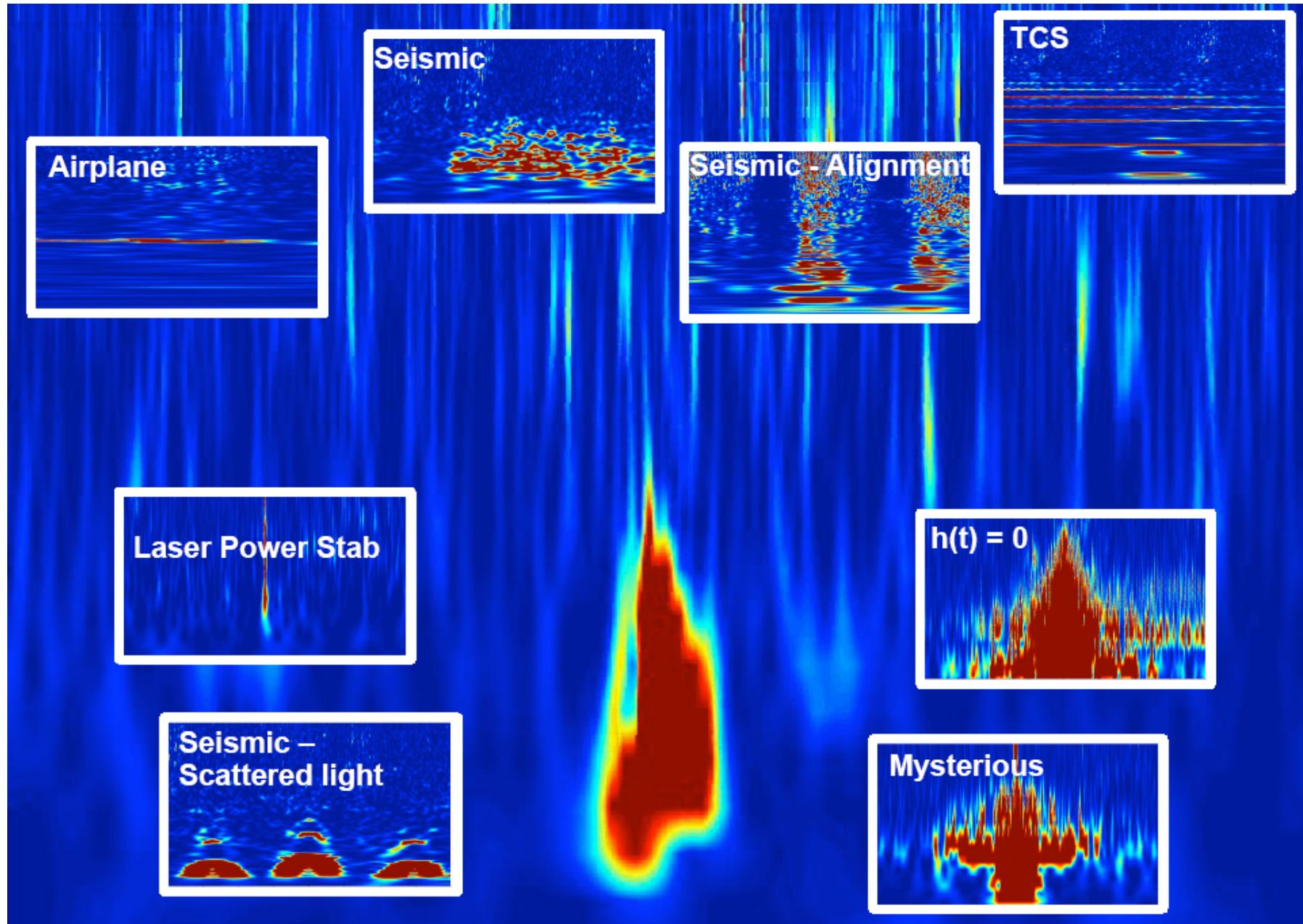
Likelihood

$$\begin{aligned} p(\mathbf{s}|\mathbf{h}) &= \prod_f \frac{1}{2\pi T S_n(f)} \exp\left(-\frac{r(f)r^*(f)}{T S_n(f)}\right) \\ &= C e^{-(r|r)/2} = C e^{-\chi^2/2} \end{aligned}$$

Noise weighted
inner product

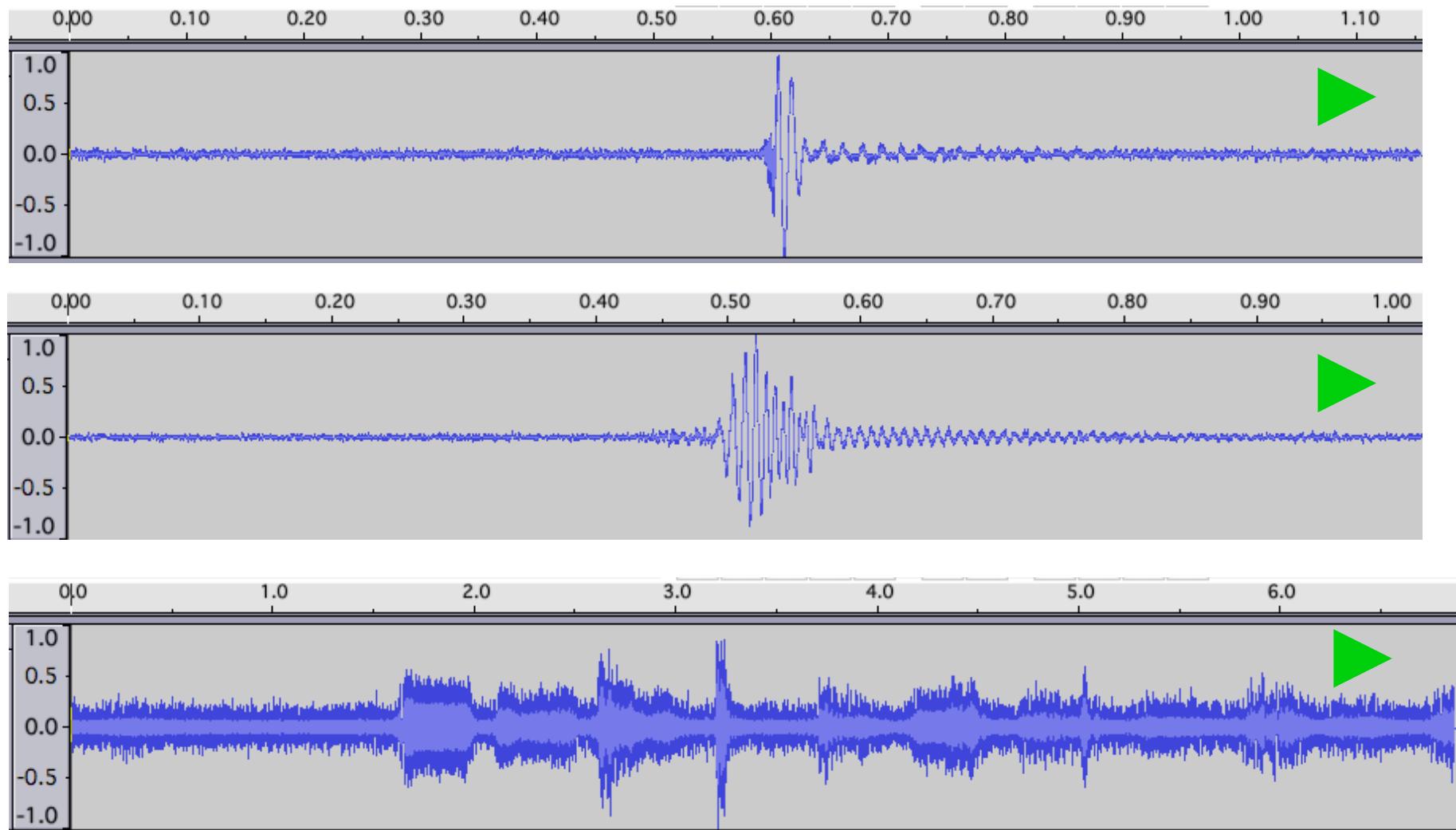
$$(a|b) = \frac{2}{T} \sum_f \frac{a(f)b^*(f) + a^*(f)b(f)}{S_n(f)}$$

Time-Frequency Scalograms of LIGO data



Time domain LIGO data

(Bandpass filtered, Whitened)

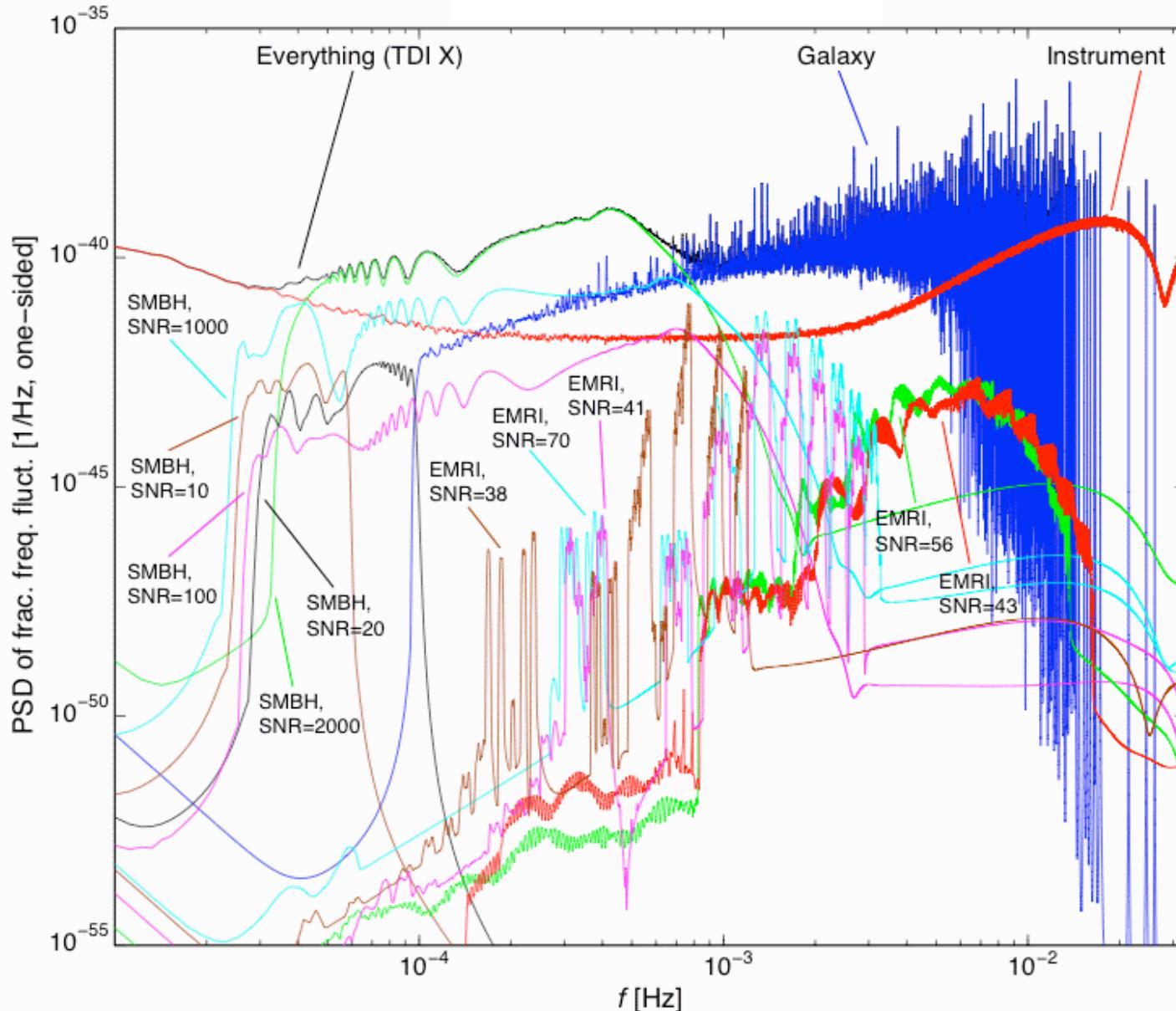


Samples from the Syracuse Audio Study of Glitches

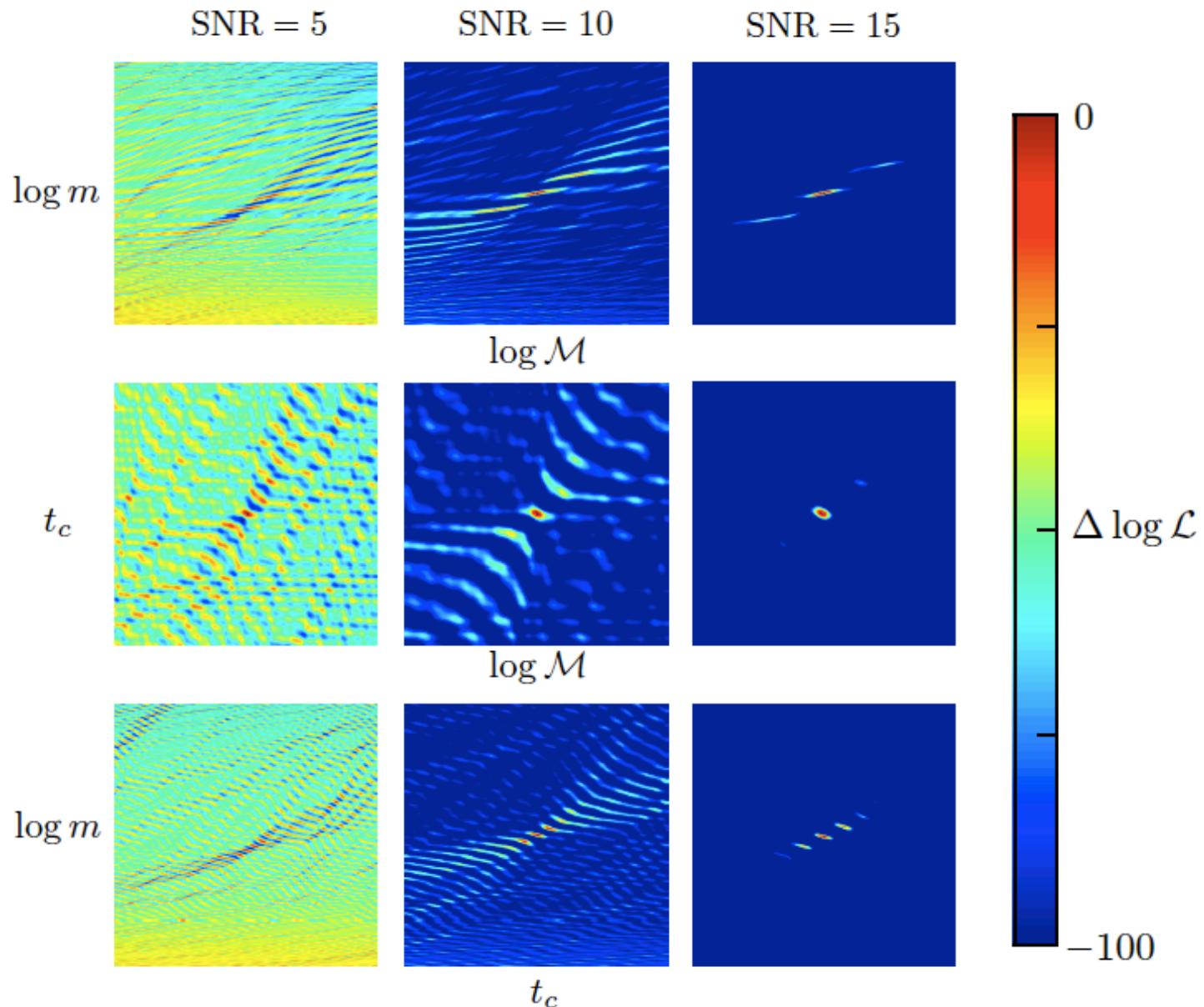
Analysis Challenges

- Rare and weak signals (terrestrial detectors)
- Non-stationary and non-Gaussian noise
- Large model dimension, small posterior to prior volume ratio
- Highly multi-modal posteriors
- Multiple overlapping signals (space based)

Multiple Overlapping Signals - Simulated LISA Data



Multi-modal Posteriors (e.g. Black Holes Inspirals)



Detecting Weak Signals

Analysis usually conducted in two stages

- Search (Optimization)
 - Maximization rather than marginalization
 - Simulated annealing, Genetic algorithms etc
- Characterization (Sampling)
 - Markov Chain Monte Carlo, Parallel tempering, DE etc
 - Nested sampling, MultiNest
- Complicated by large model dimension, small posterior to prior volume ratio, e.g. $\frac{\Delta V_{90}}{V} \sim 10^{40}$
- Can exploit multi-modality of posterior

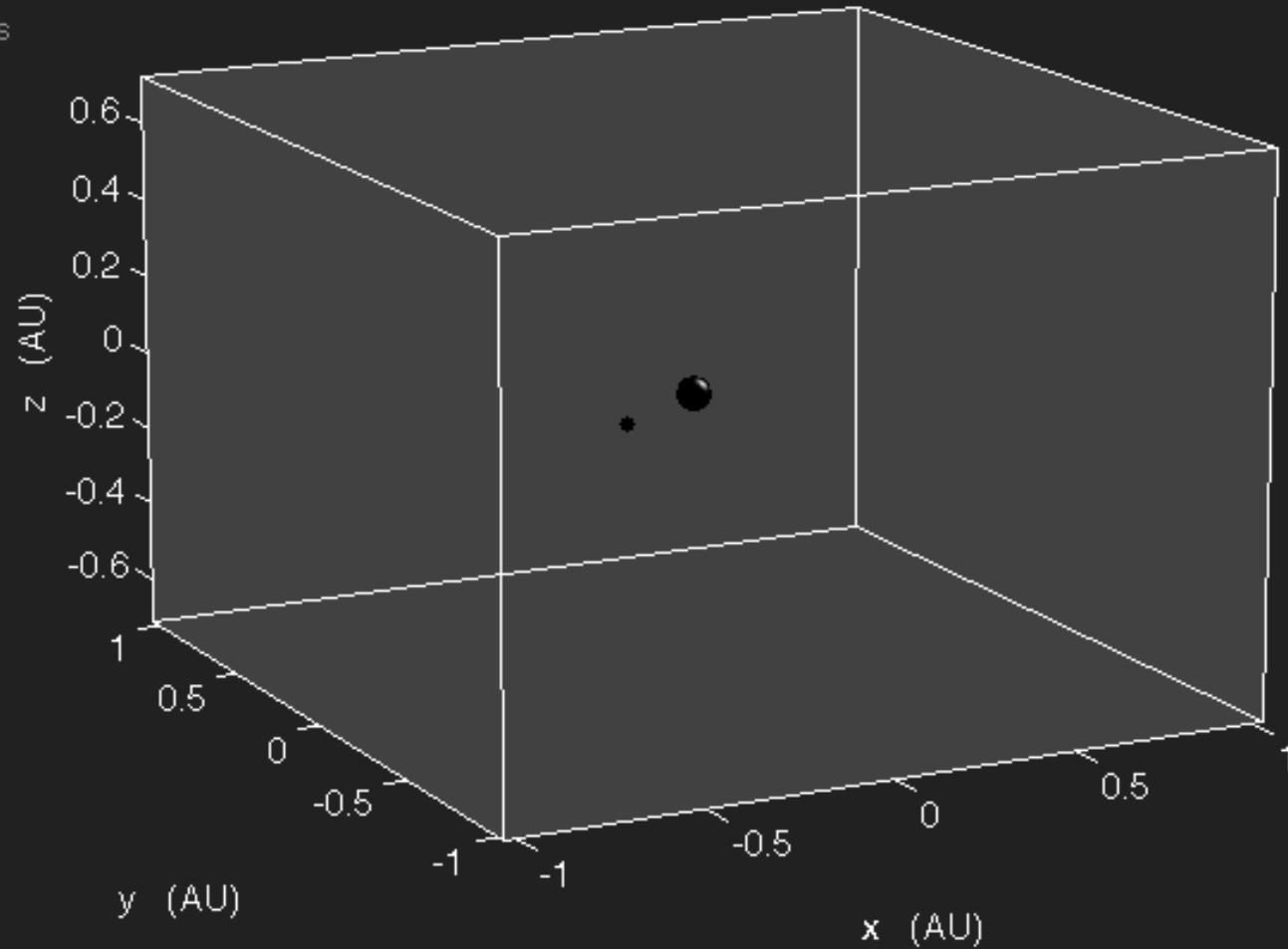
Extreme Mass Ratio Inspirals

Large black hole:
shown to scale
3,000,000 solar masses
90% maximal spin

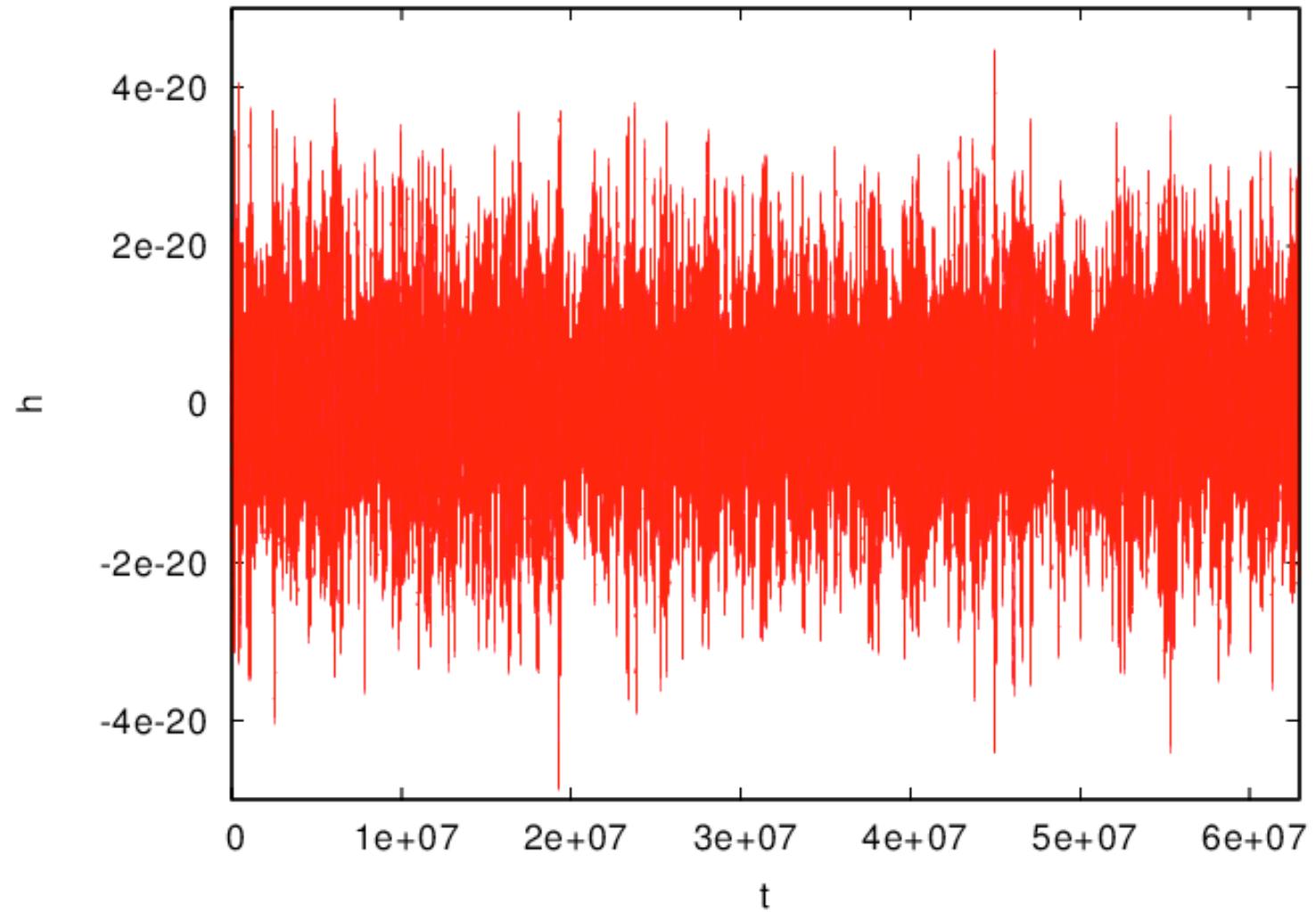
Small black hole:
shown enlarged
540 solar masses
negligible spin

Trace duration:
1 day

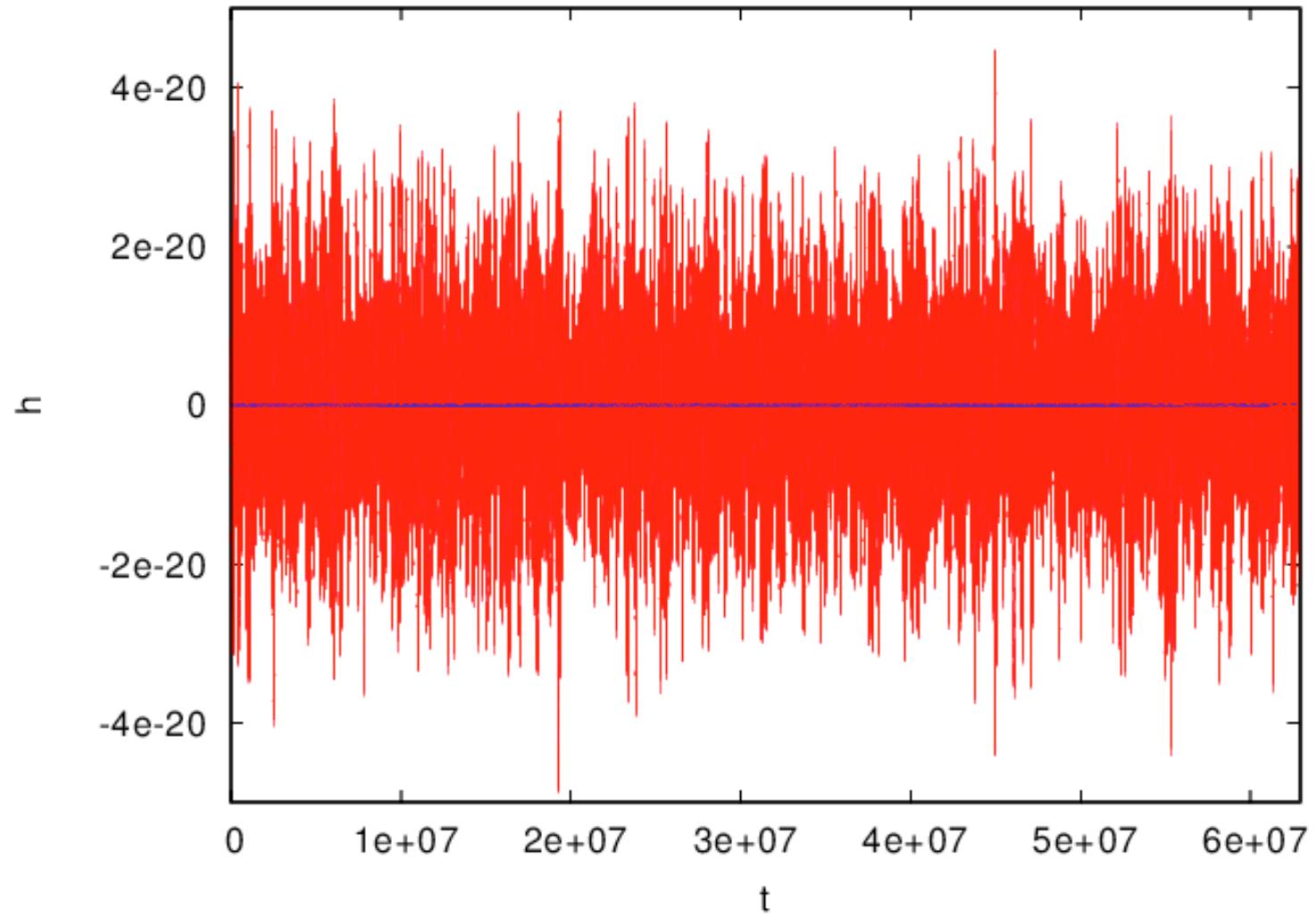
Steve Drasco
Max Planck Institute
for Gravitational Physics
(Albert Einstein Institute)
sdrasco@aei.mpg.de



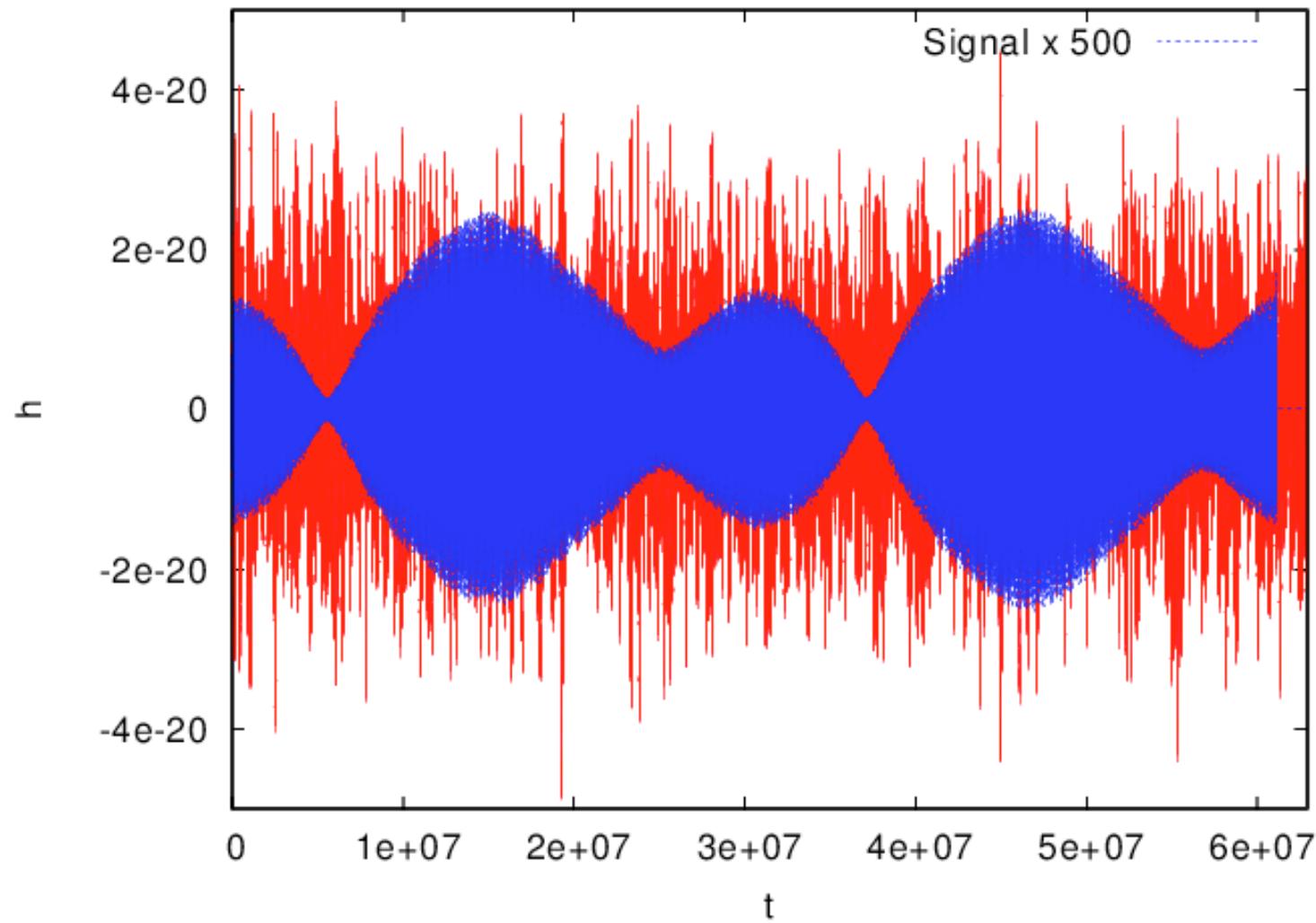
Finding weak signals



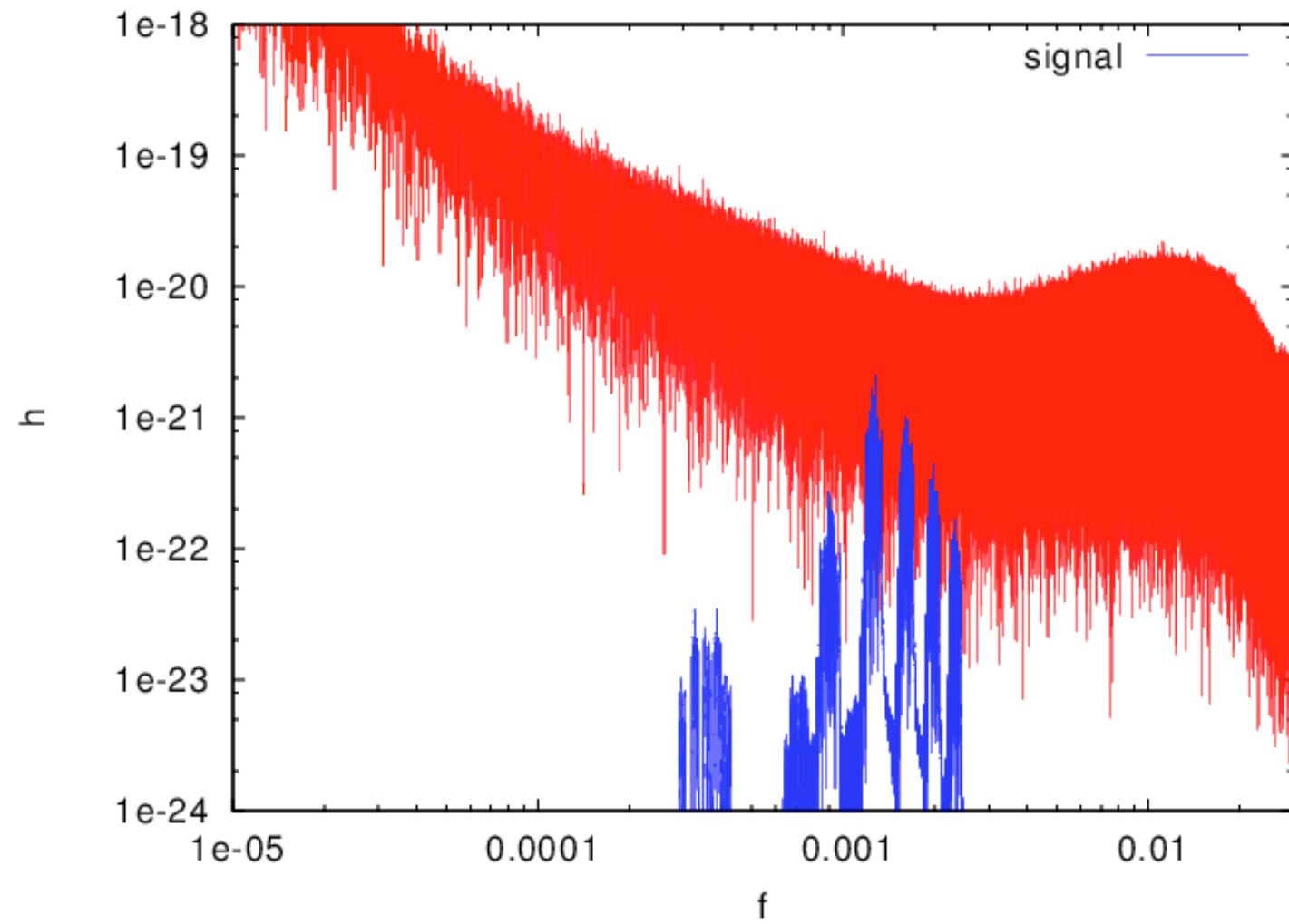
Finding weak signals



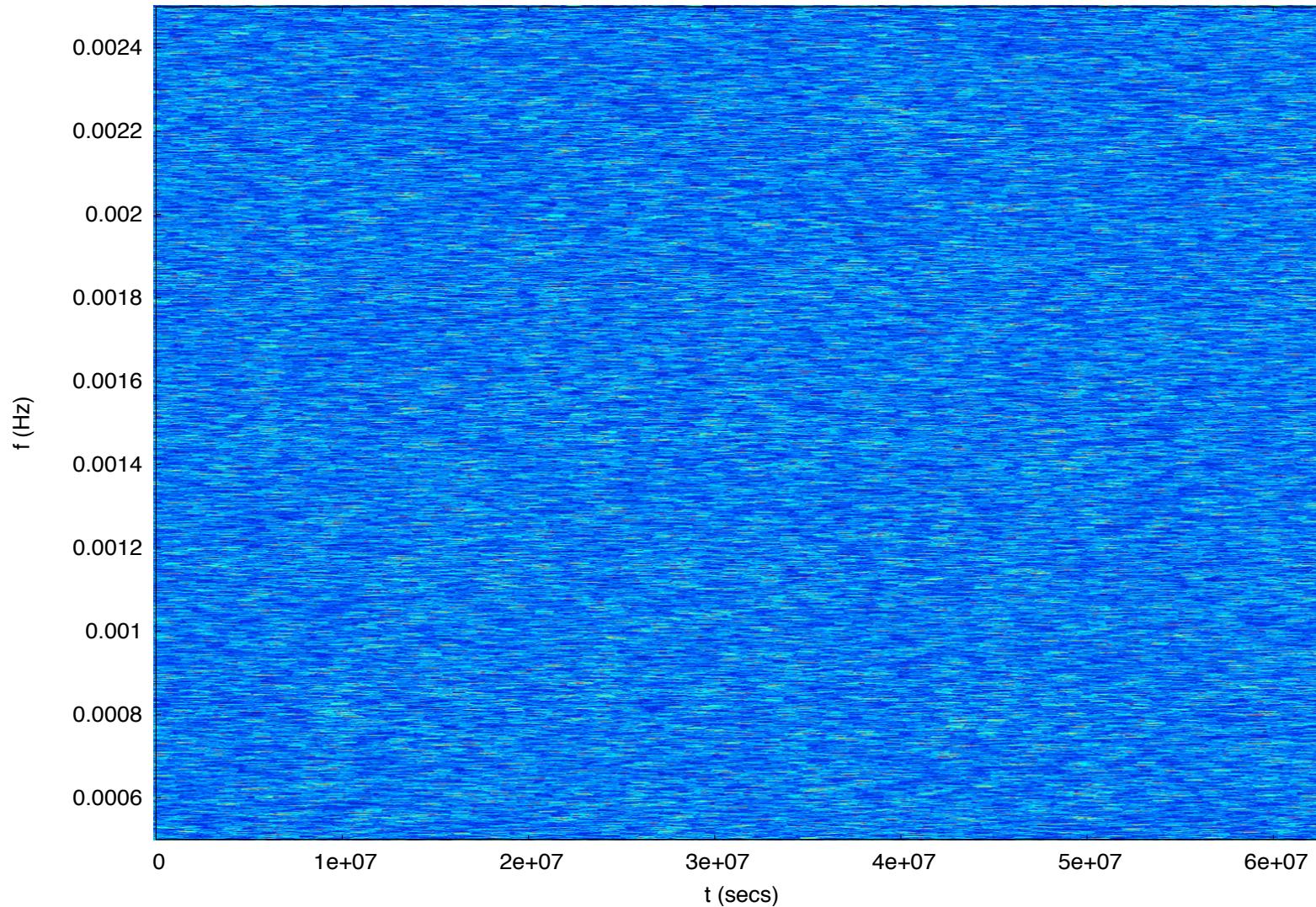
Finding weak Signals

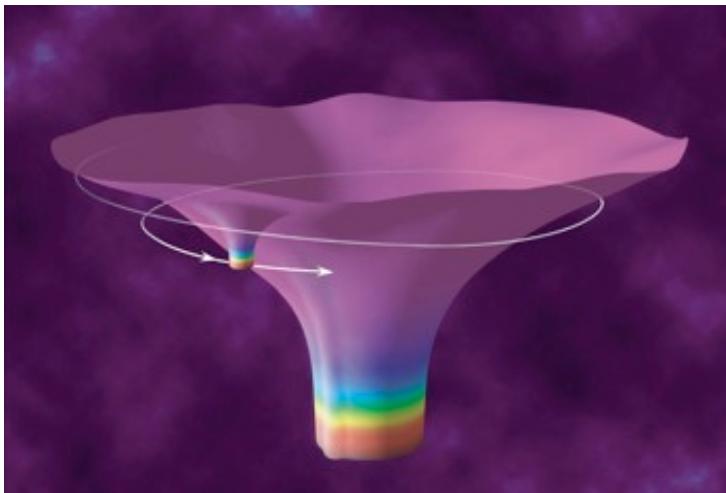


Easier to see in Frequency?



Easier to see in Time-Frequency?





Mock LISA Data Challenge

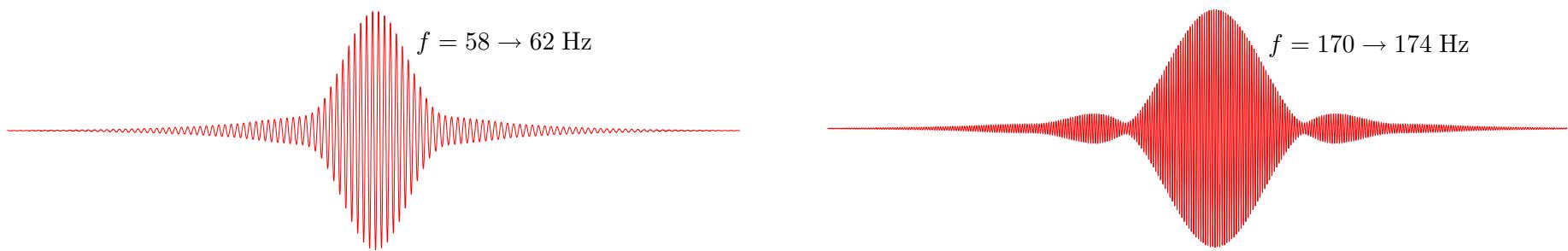
N. J. Cornish, Class. Quant. Grav. 28, 094016 (2011)

SNR = 21.6

SourceType	ExtremeMassRatioInSpiral	ExtremeMassRatioInSpiral
EclipticLatitude	0.421812762567	0.428327994937
EclipticLongitude	6.12722883835	6.086378539
Polarization	0	0
PolarAngleOfSpin	1.12926516708	1.06896570232
AzimuthalAngleOfSpin	1.02683301361	1.050521566
Spin	0.607839950171	0.6078238235
MassOfCompactObject	10.3529155149	10.35303219
MassOfSMBH	9592613.64626	9593171.82
InitialAzimuthalOrbitalFrequency	0.000195874494514	0.000195874648
InitialAzimuthalOrbitalPhase	5.21945959065	5.270798052
InitialEccentricity	0.181454715223	0.181500634
InitialTildeGamma	5.87897706737	2.707804463
InitialAlphaAngle	4.97256712568	4.908907739
LambdaAngle	0.809747003385	0.809309222072
Distance	1173313438.97	1022400436.0

Developing a realistic noise model: BayesWave

- Bayesian model selection
 - Three part model $\mathbf{s} = \mathbf{R} \cdot \mathbf{h} + \mathbf{g} + \mathbf{n}$
 - Trans-dimensional Markov Chain Monte Carlo
- Wavelet decomposition
 - Glitch model parameters are wavelet amplitudes
 - Number, amplitude and location of “active” pixels varies



T. B. Littenberg, N. J. Cornish, Phys. Rev. D82, 103007 (2010).

Three part model

Noise model n - defines the likelihood function

$$p(s|a_{ij}) = \frac{1}{\sqrt{2\pi}\sigma_i} e^{-\frac{(s_{ij}-a_{ij})^2}{2\sigma_i^2}}$$

Ideally would also include information from the auxiliary channels

Noise level also an unknown,
determined from data

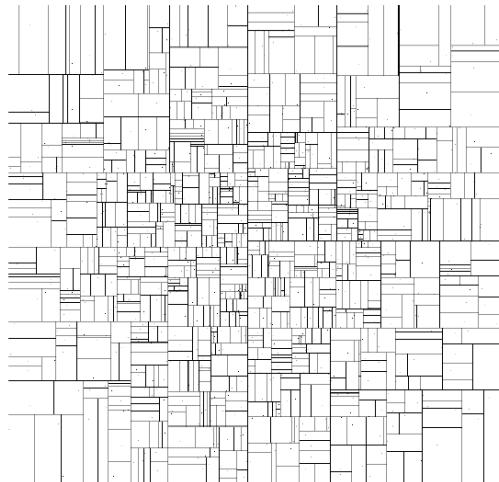
Glitch model g - clustered TF power, incoherent across network

Signal model h - analytic waveforms or clustered TF power, coherent across network
- can incorporate priors such as elliptical polarization

Transitions between signal+noise and noise only models

- Very difficult to compute the evidence for each model individually due to variable and large dimensionality of the glitch model
- Difficult to jump to the signal model since $\Delta V_{90}/V \ll 1$

One solution: Use PDFs from each model as proposal densities

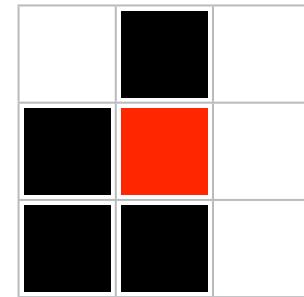


PDFs can be mapped by sparse matrix representations or KD trees

T. B. Littenberg, N. J. Cornish, Phys. Rev. D82, 103007 (2010).

W. M. Farr and I. Mandel, arXiv:1104.0984 [astro-ph.IM] (2011).

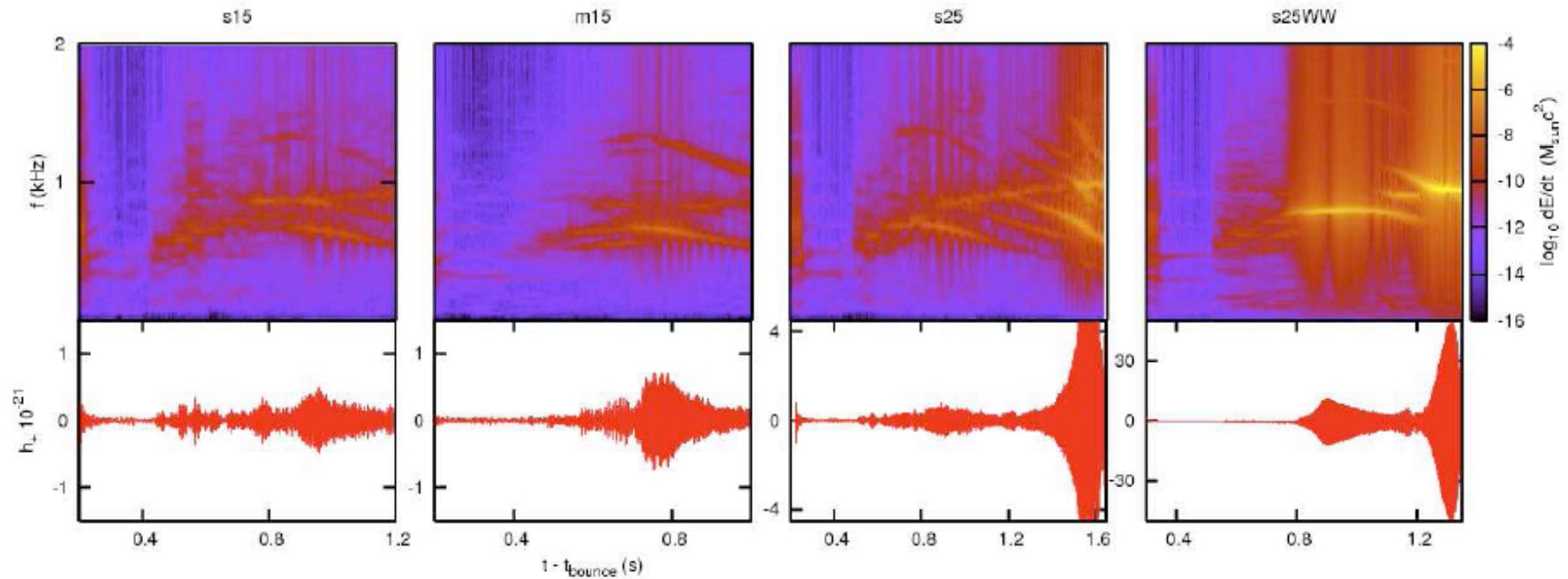
Priors

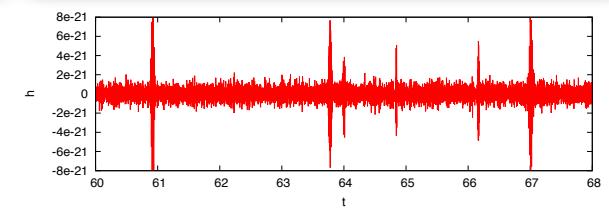
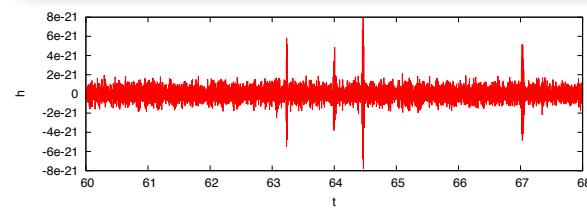
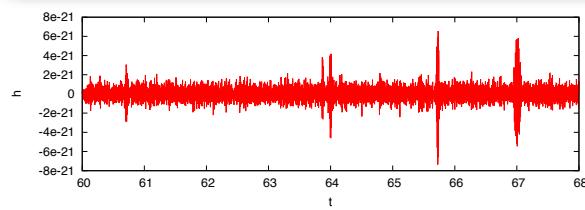
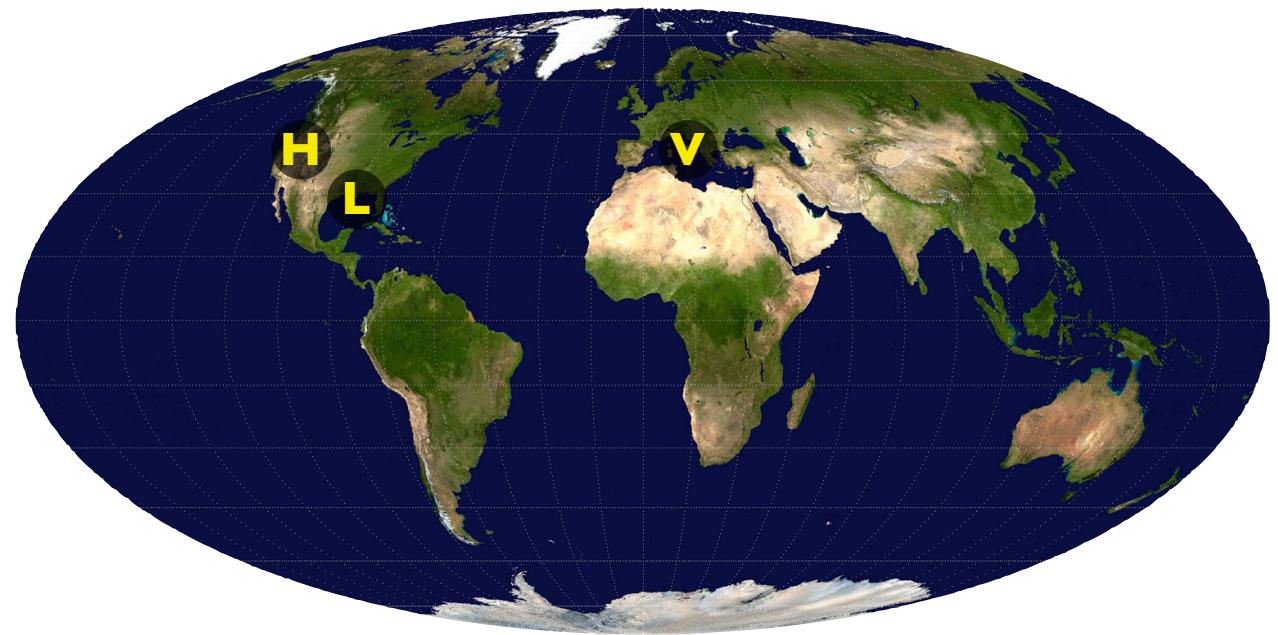


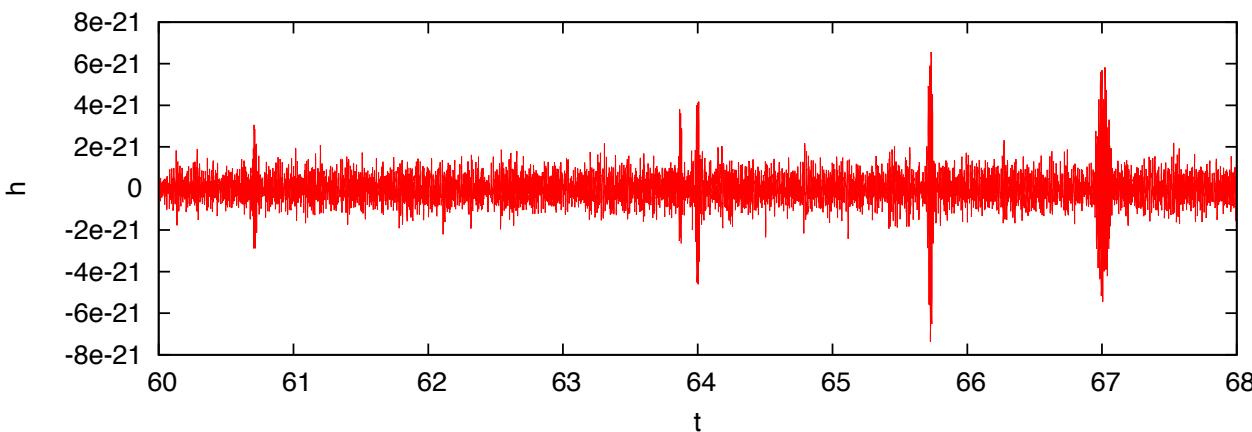
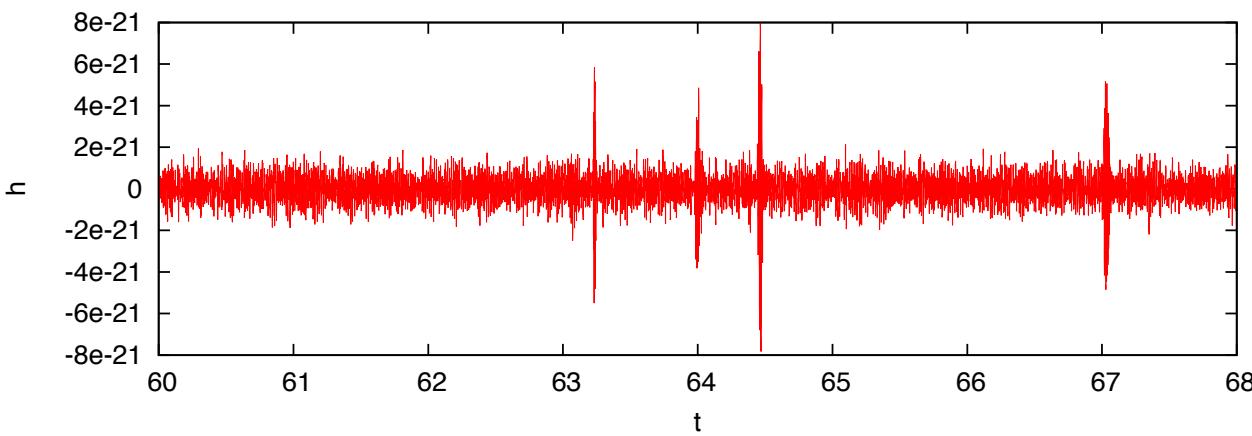
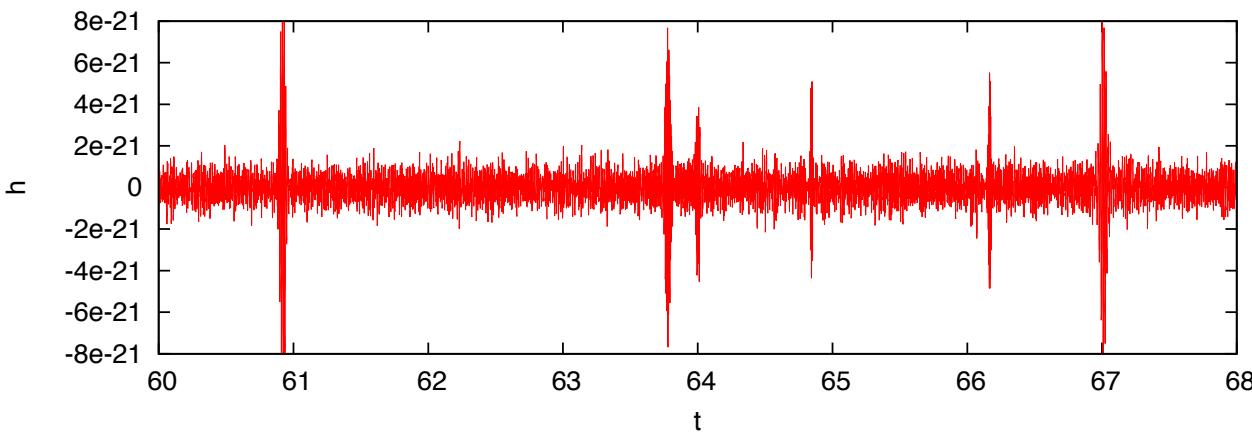
Clustering Prior: $p(ij) \propto \# \text{nearest neighbours}$

Amplitude Prior: $p(a_{ij}) = \frac{1}{\sqrt{2\pi}\kappa\sigma_i} e^{-\frac{a_{ij}^2}{2\kappa\sigma_i^2}}$

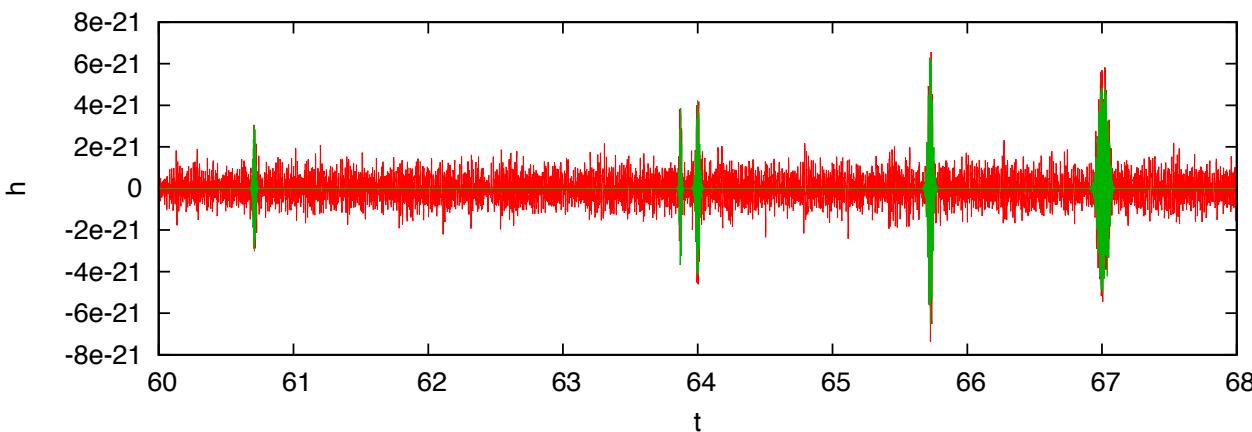
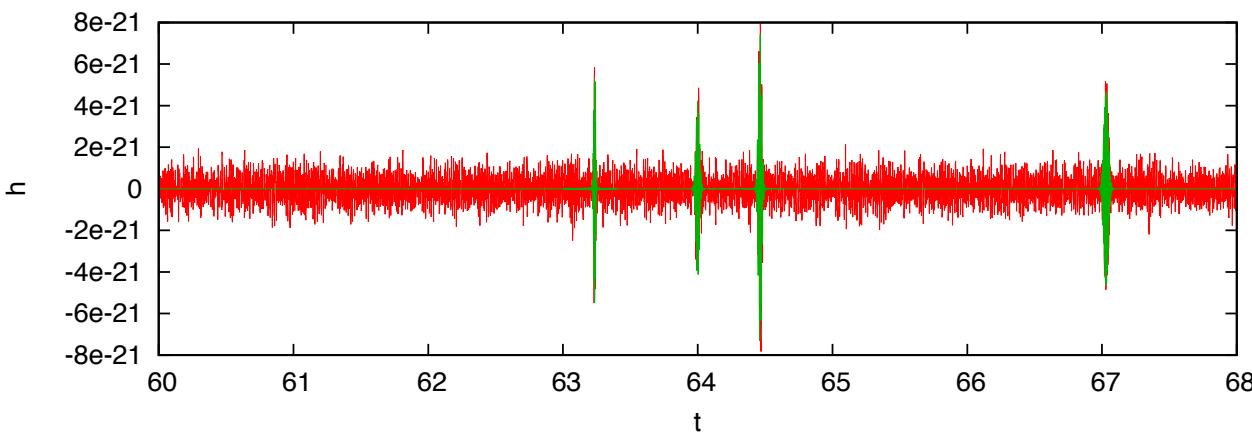
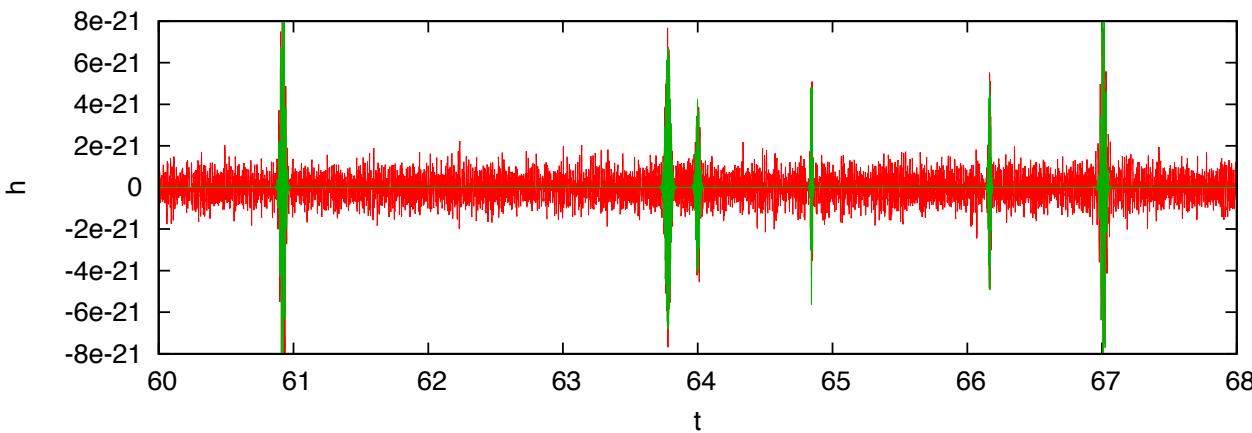
Astrophysical model Priors: e.g. time-frequency content of a core collapse SN



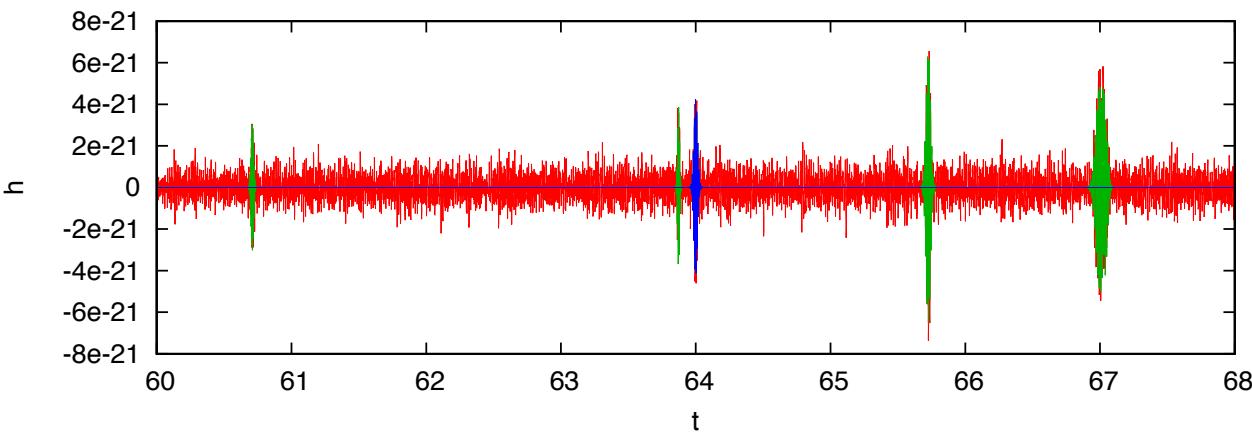
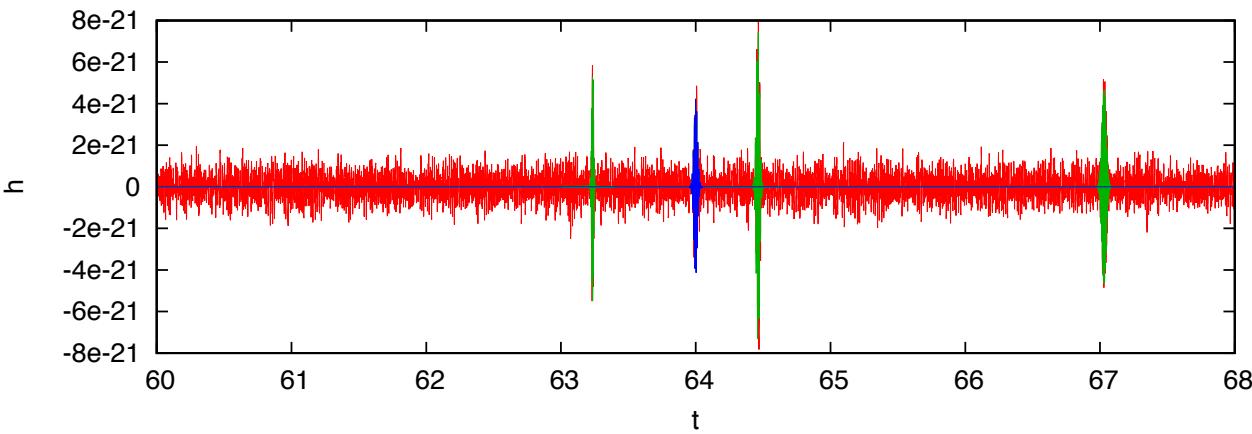
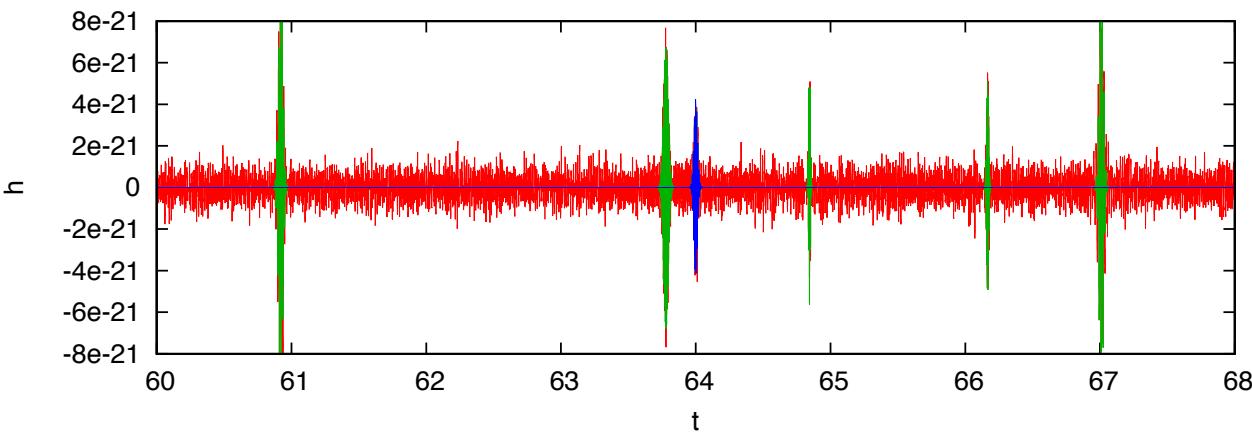


H**L****V**

Simulated data, band-passed 60-300 Hz

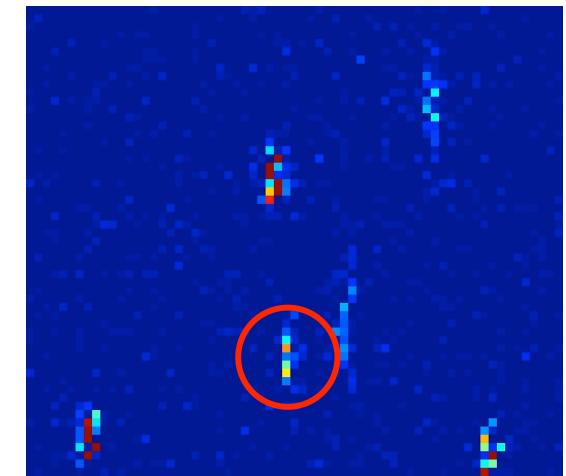
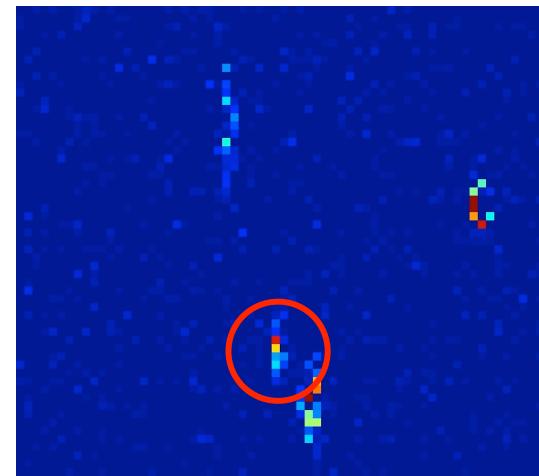
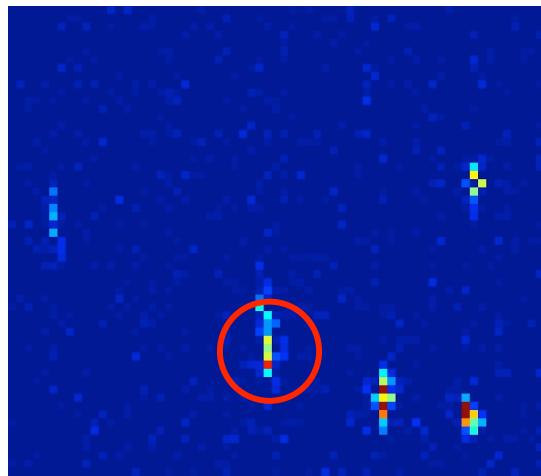
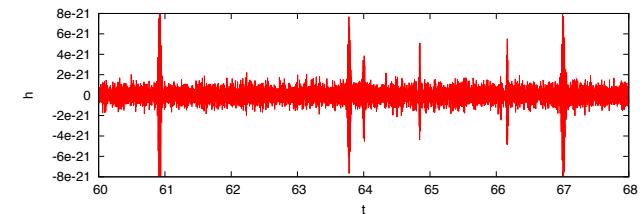
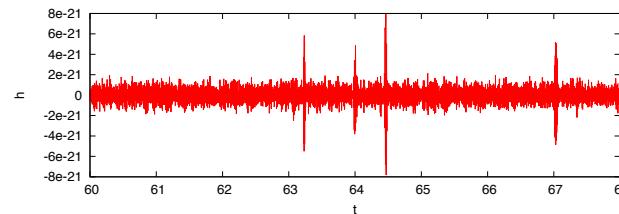
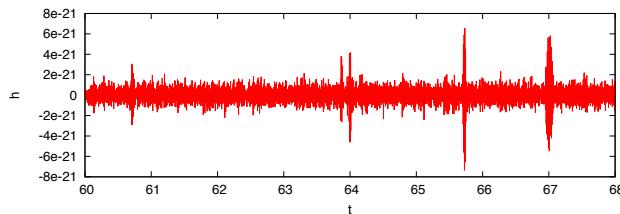
H**L****V**

Simulated data, band-passed 60-300 Hz

H**L****V**

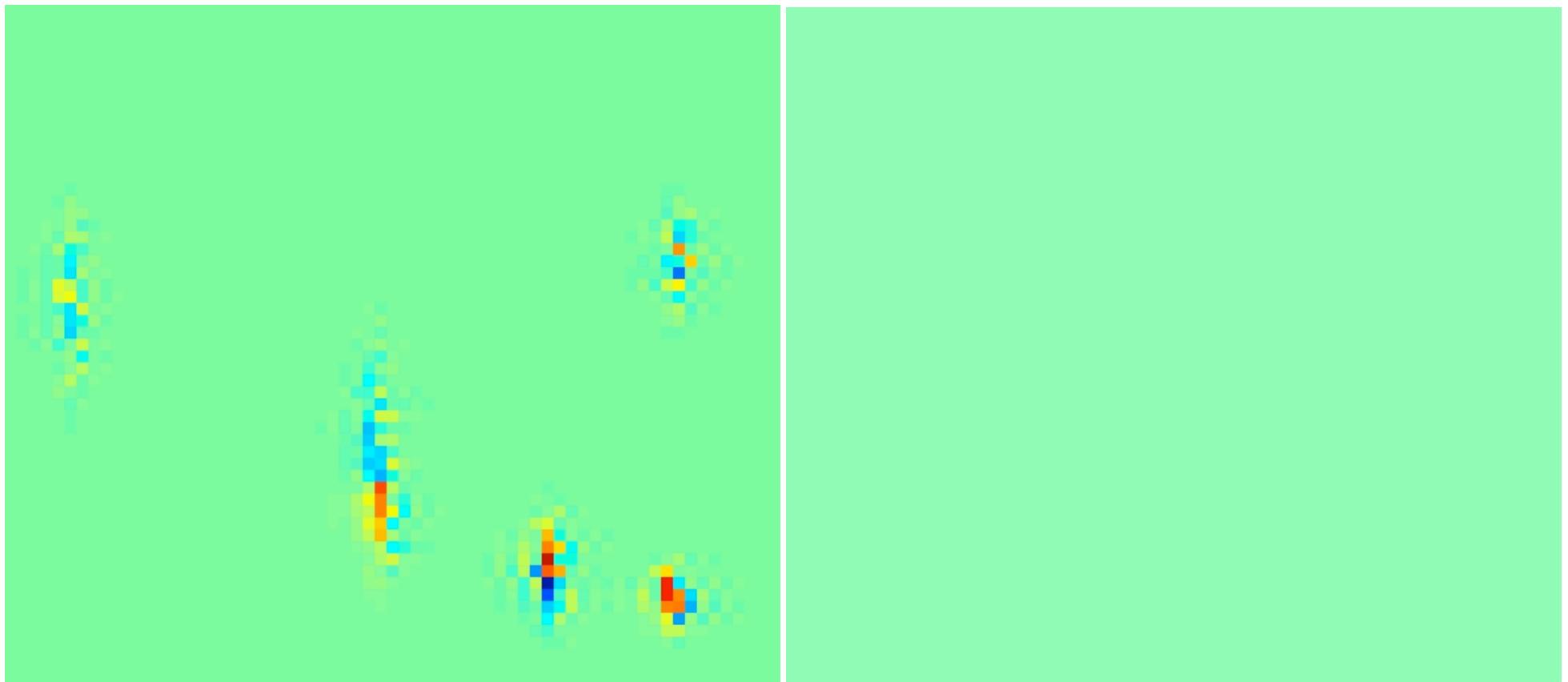
Simulated data, band-passed 60-300 Hz

Coherent Network Analysis



GW signals line up in time *and* frequency
and morphology

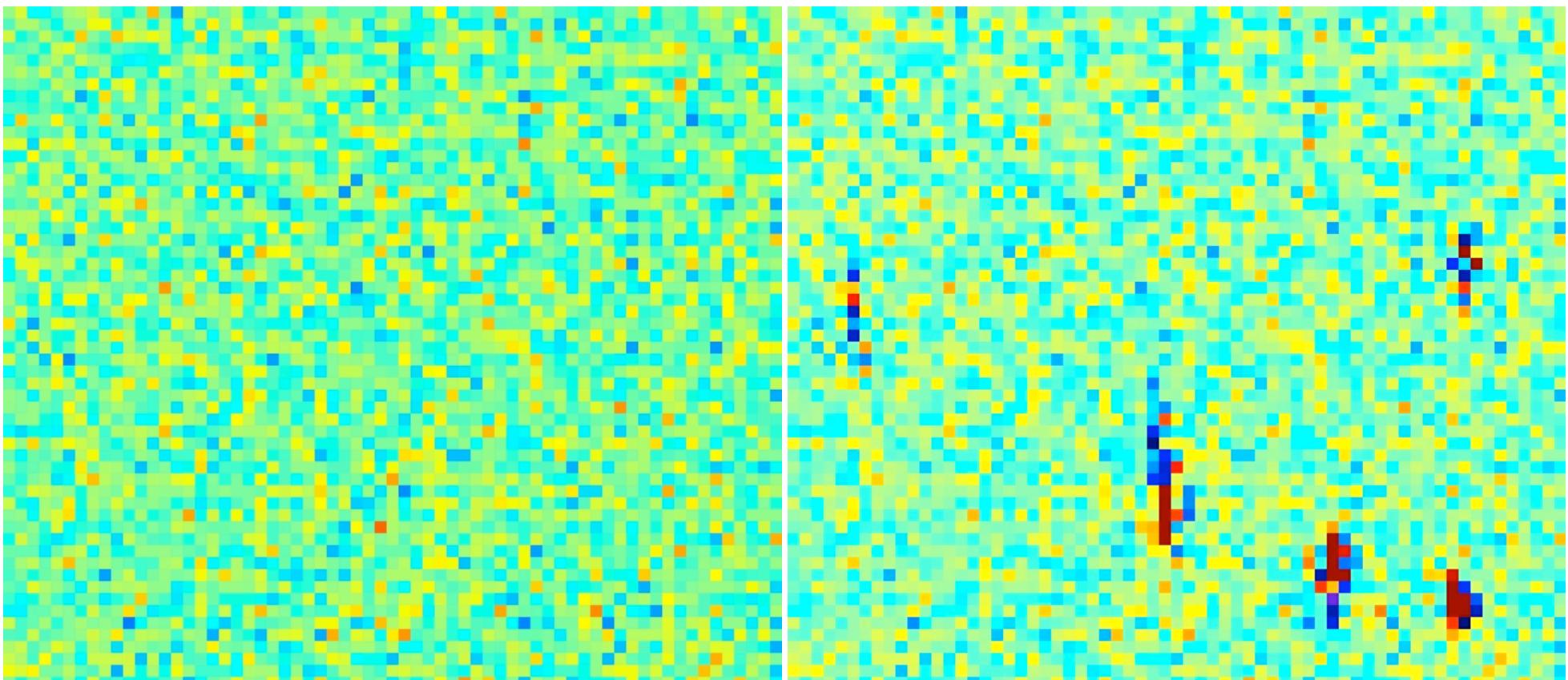
Glitch & Signal Recovery



Injected

Recovery

Noise Recovery



Injected

Recovery

Help Wanted

- First direct detection of gravitational waves within 5 years
- Terabytes of data, weak signals, complex waveforms, glitchy detectors

...and a bunch of theoretical physicist trying to do the analysis