

Data Analysts Without Borders: A Case Study



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Warren G. Anderson
University of Wisconsin - Milwaukee



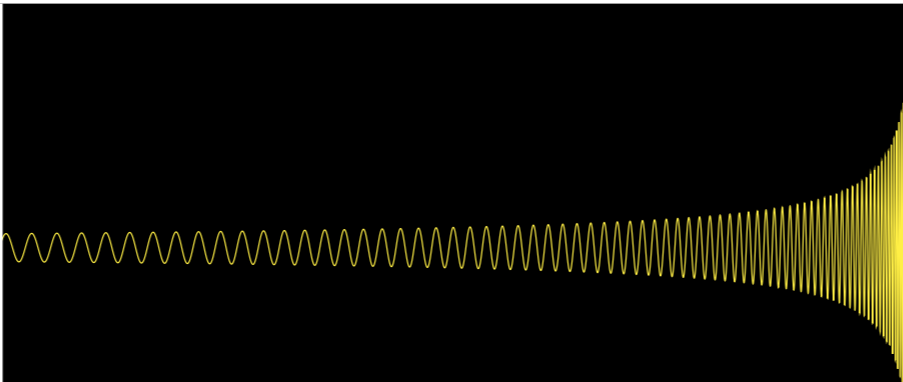
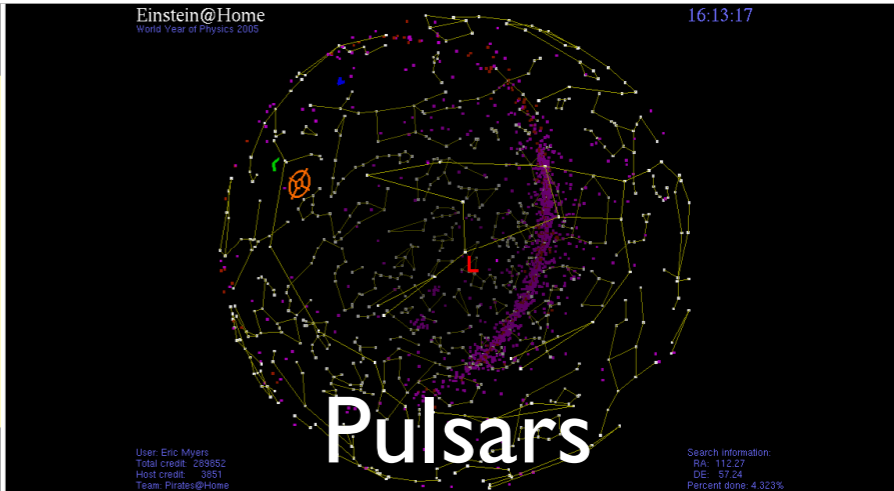
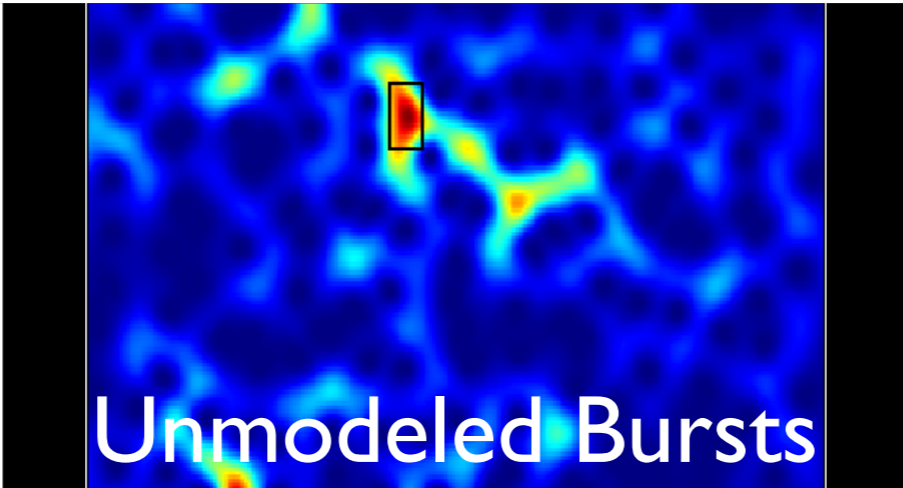
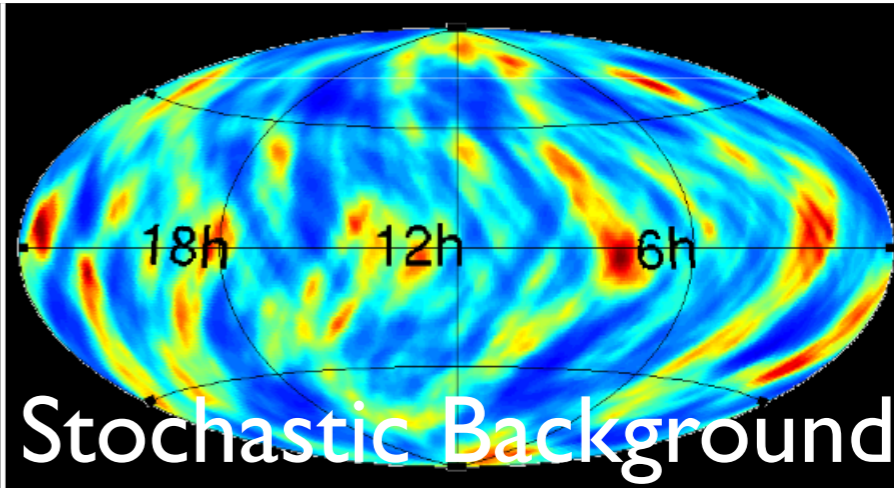
Image courtesy of NASA
Goddard NR group

Outline

- Conventions
 - Four types of DA
- Foundations
 - Data, Likelihood, Power, Power²
- STAMP
 - Motivation, Idea, Progress

Convention

- Ground-based gravitational wave DA is often divided into four categories:

Category	Short Duration	Long Duration
Theoretical Waveform	 <p data-bbox="900 1266 1495 1360">Binary Inspirals</p>	 <p data-bbox="1970 1266 2252 1360">Pulsars</p>
No Theoretical Waveform	 <p data-bbox="831 1759 1561 1854">Unmodeled Bursts</p>	 <p data-bbox="1668 1759 2554 1854">Stochastic Background</p>

Data

- From our instruments we get strain data $h(t)$

$$h(t) = n(t) + s(t)$$

Noise *Signal*

Data

- From our instruments we get strain data $h(t)$

$$h_i = n_i + s_i$$

Noise *Signal*

Data

- From our instruments we get strain data $h(t)$

$$h_i = \underbrace{n_i}_{\text{Noise}} + \underbrace{s_i}_{\text{Signal}}$$

- Signal is deterministic (possibly $s_i = 0$)
- Noise is stochastic - eg Gaussian (ideally)

$$p(\mathbf{n}) \propto \exp\left(-\frac{\mathbf{n} \cdot \mathbf{n}}{2\sigma}\right)$$

Likelihood

- Neyman-Pearson: Optimal statistic is **Likelihood**

$$\Lambda[h] = \int \frac{p(h|\mathbf{n} + \mathbf{s})}{p(h|\mathbf{n})} \mathcal{D}[\mathbf{s}]$$

- The measure $\mathcal{D}[\mathbf{s}]$ specifies the class of signals to search for by projecting, eg to search for a single specific signal $\hat{\mathbf{s}}(t)$, let $\mathcal{D}[\mathbf{s}] = \delta(\mathbf{s} - \hat{\mathbf{s}}) d\mathbf{s}$.

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Signal

- Probabilities of h derive from distribution of \mathbf{n} , eg

$$p(h|\mathbf{n} + \mathbf{s}) \propto e^{-\frac{(\mathbf{h} - \mathbf{s}) \cdot (\mathbf{h} - \mathbf{s})}{2\sigma}}$$

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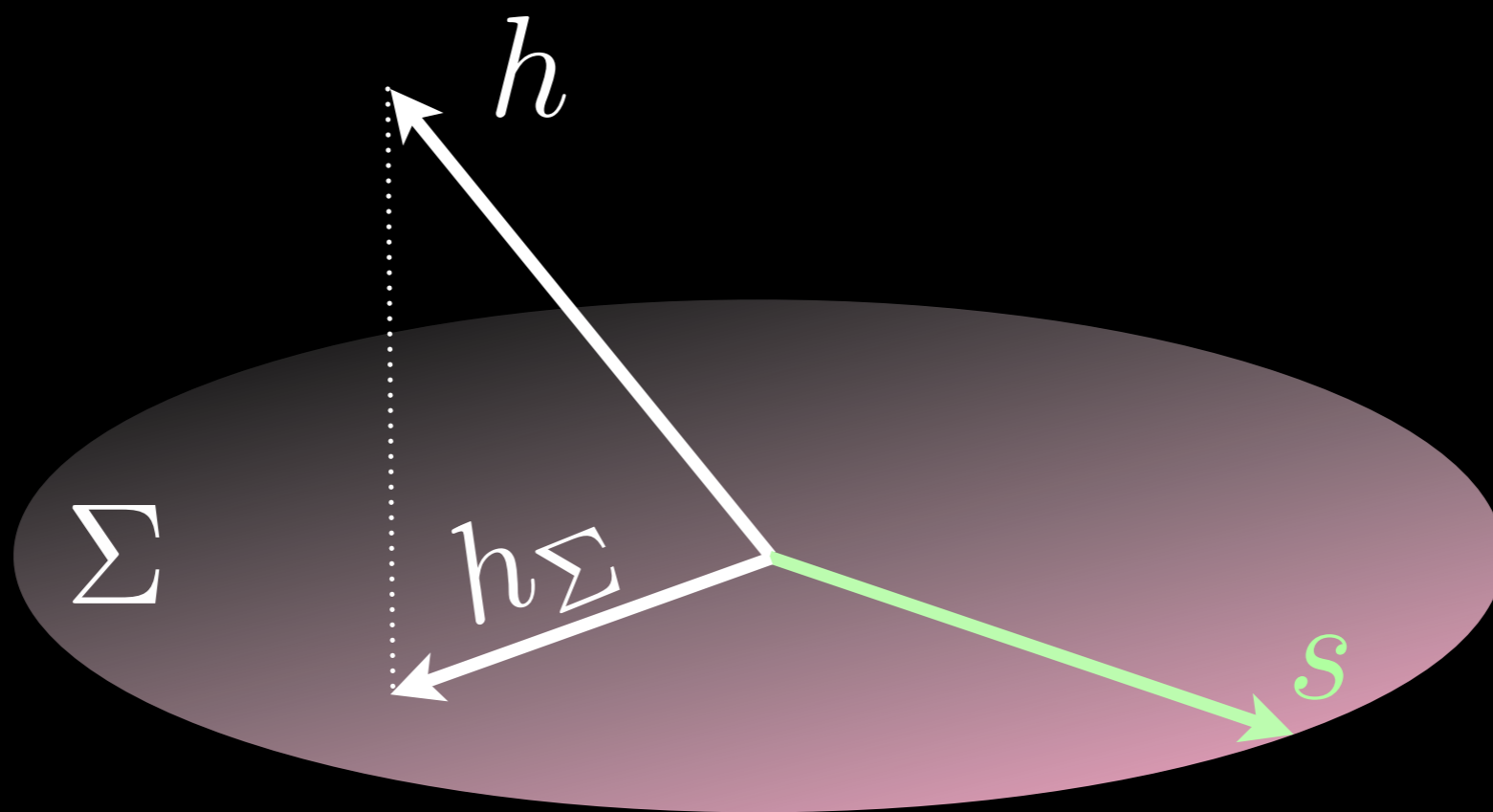
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Power

- Putting this together for Gaussian noise:

$$\Lambda \propto \int_{\Sigma} \exp(h_{\Sigma} \cdot \mathbf{s} - \mathbf{s} \cdot \mathbf{s}/2) d\mathbf{s}$$



Power

- Putting this together for Gaussian noise:

$$\Lambda \propto \int_{\Sigma} \exp(h_{\Sigma} \cdot s - s \cdot s/2) ds$$

- Note that Λ is monotonic in length of projection, $|h_{\Sigma}|$.
- **Power** of projected data, $|h_{\Sigma}|^2$, is equivalently optimal statistic.

Power²

- For more than one detector, simply substitute:

$$h = A^i(\Omega)h_i(t_i)$$

- Eg, for two detectors with Gaussian noise, optimal statistic is

$$\begin{aligned} |h_\Sigma|^2 &= | [A_1 h_1 + A_2 h_2]_\Sigma |^2 \\ &= A_1^2 |h_1^2|_\Sigma + A_2^2 |h_2^2|_\Sigma \\ &\quad + 2A_1 A_2 [h_1 \cdot h_2]_\Sigma \end{aligned}$$

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Glitches

- Real interferometer noise is not Gaussian, but has “glitches”.
- Two problems:
 - No analytic model - need time-slides or other methods to assess false alarm/detection stats.
 - Power statistic no longer optimal because loud glitches in one instrument cause large auto-power but real signals should have cross-power.

Idea

- Use cross-power to look for various types of signals.
- Advantages:
 - Cross-power statistics very close to Gaussian - close enough?
 - Cross-power statistic less susceptible to glitches.

STAMP

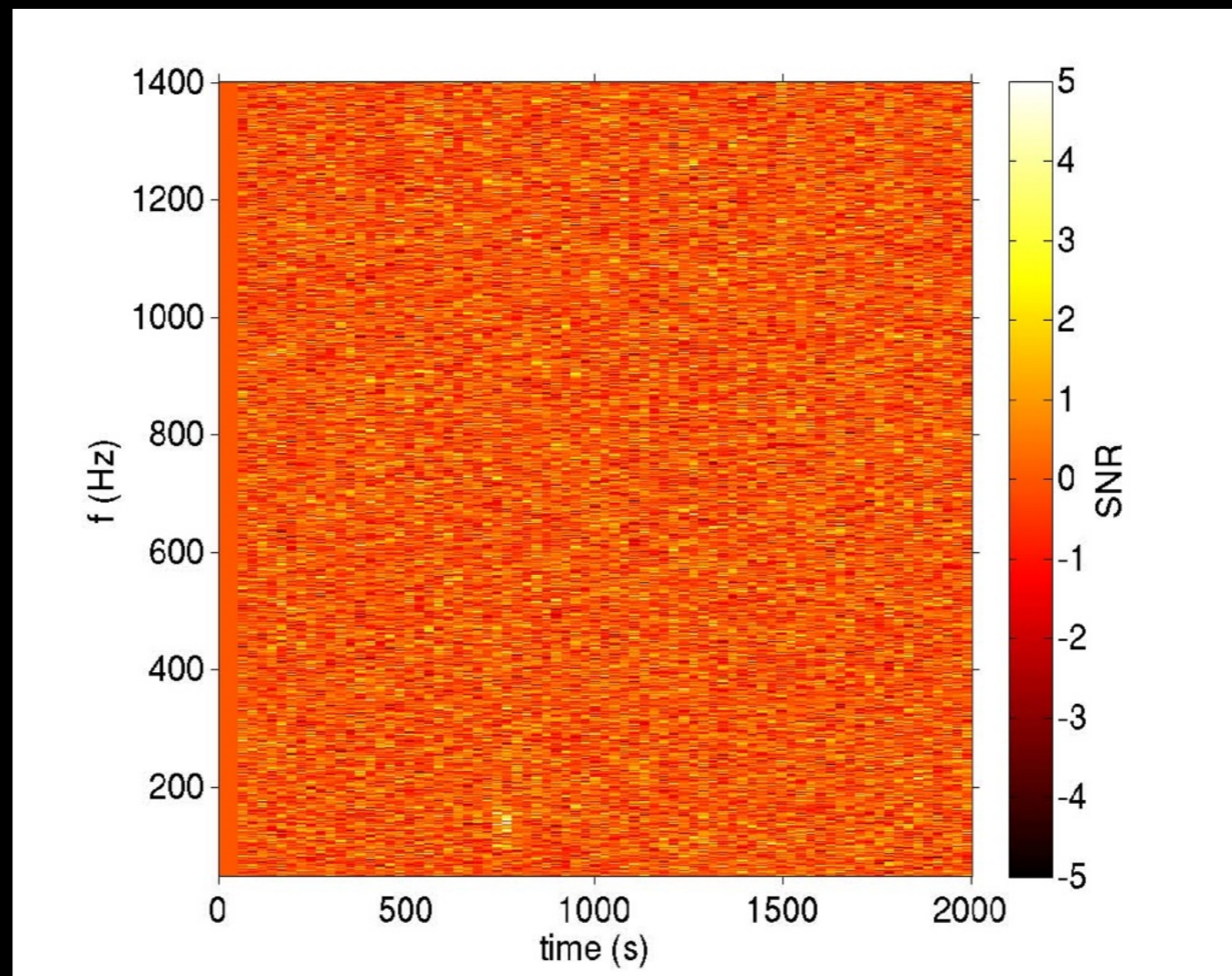
- Stochastic Transient Analysis Multi-detector Pipeline
- Participants: N. Christensen, M. Coughlin, S. Dorsher, S. Giampanis, S. Kandhasamy, V. Mandic, A. Mytidis, C. Ott, T. Prestegard, P. Raffai, E. Thrane, B. Whiting, WGA.
- Astrophysical searches for longer-lived transients (seconds - weeks) in LIGO-Virgo data.
- Searches for transients with various degrees of prior knowledge - serendipitous to targeted.
- Also useful for analyzing noise monitoring channels.

Tools

- Cross-power data
 - used in stochastic analysis - data and expertise available.
- Box search, radon transform, locust search:
 - used in unmodeled burst searches - codes and expertise available.
- Hough search:
 - used for pulsar and burst searches - codes and expertise available.

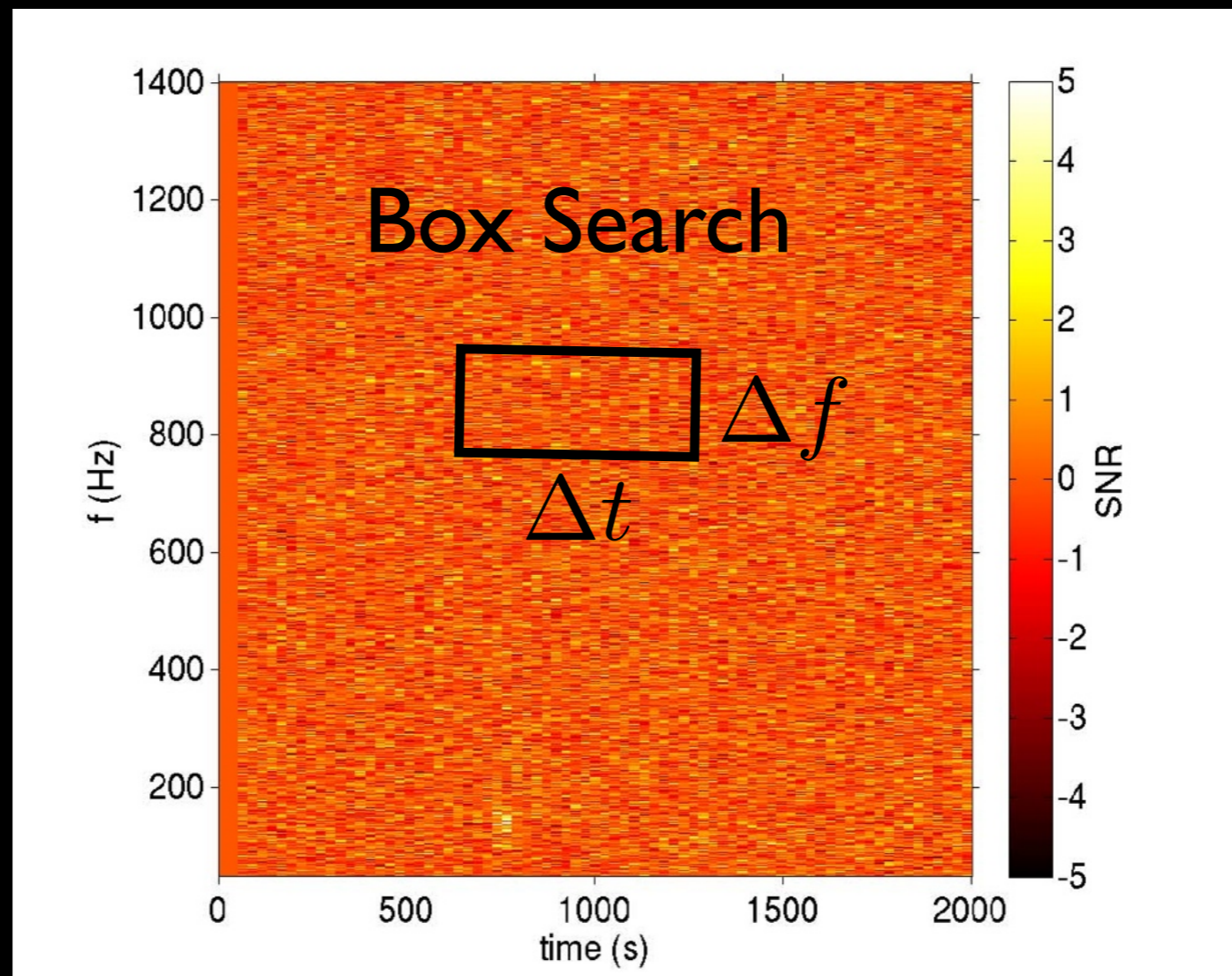
Maps

- We use a TF representation of cross-power to make projection onto signal space easier.



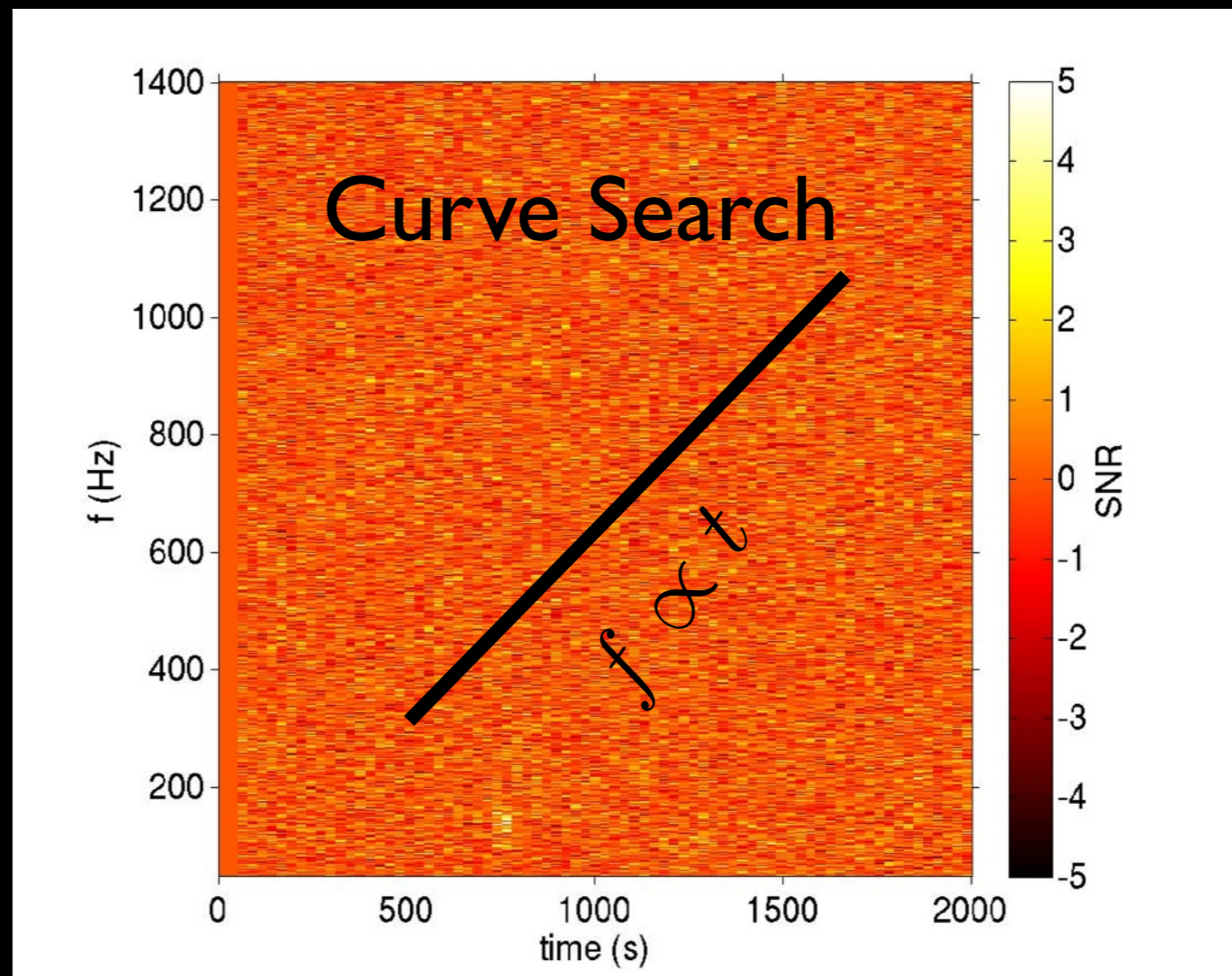
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Activity

- Astrophysical sources: Ott, Giampanis, Thrane
- Statistics of cross-power: Mytidis, Whiting
- Efficiency of cross-power: Giampanis, Thrane, WGA
- Searches for astrophysical signals: Dorsher, Kandhasamy, Mandic, Thrane, Christensen, Coughlin, Dorsher, Giampanis, Prestegard, Raffai
- Noise studies: Christensen, Coughlin, Thrane
- Writing methods paper: all

Conclusions

- All “four types” of LIGO-Virgo data analysis have a lot in common:
 - use the same data
 - based on likelihood
 - the only difference is signal space we project on.
- Expertise and tools are portable across many analyses.
- STAMP is proving to be an example of fruitful interactions between analysts from different camps.

Sources

- core-collapse SN / long GRBs: protoneutron star turbulence, rotational instabilities (bar modes, r-modes), accretion disk instabilities (excitations, disintegration)
- binary inspirals: remnant protoneutron star turbulence, rotational instabilities in remnant, highly-eccentric binary black holes
- isolated neutron stars: pulsar glitches, quasi-periodic oscillations after SGR flares
- other?