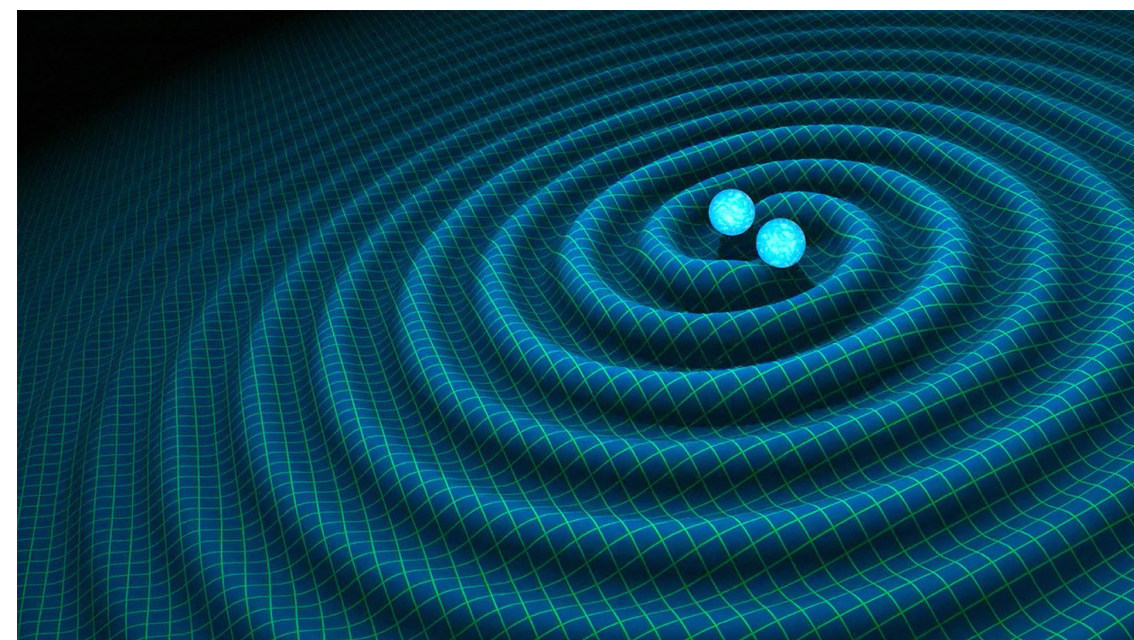
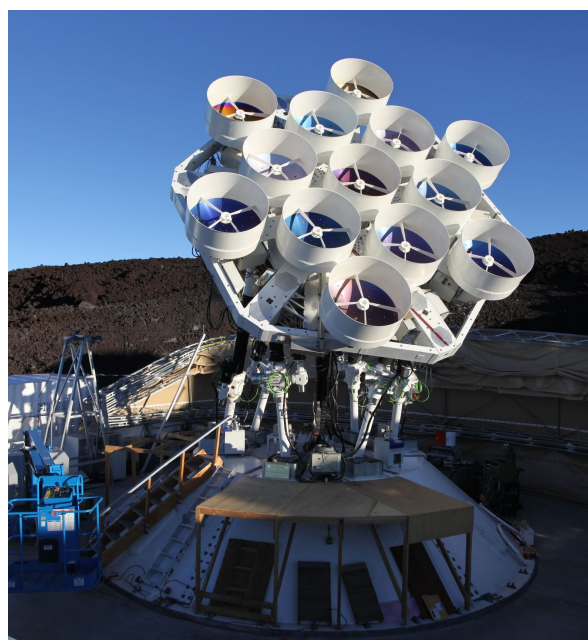


Detection of gravitational waves and data pipeline for compact binary coalescences and stochastic gravitational wave background

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2019 APCTP School on Gravitational-Wave Cosmology



Data analysis groups in Taiwan

- Feng-Li Lin's group @ National Taiwan Normal University
 - CBC Search pipeline with GPU
 - CBC search/parameter estimation with deep learning
 - Parameter estimation
- Cosmology group @ Tamkang University
 - Theoretical studies on boson star
 - Data analysis of stochastic gravitational wave background (SGWB)

Sources of LIGO-VIRGO-KAGRA

- Compact Binary Coalescence (CBC)
 - BNS, BBH, BH-NS..
- Burst: poorly modeled signal
 - Core-collapse supernovae
 - Long Gamma Ray Burst
 - Cosmic string
- Continuous Waves: long lived, nearly sinusoidal and weak
 - Rotating neutron stars
- Stochastic Background
 - Cosmology: inflation, cosmic string
 - Astrophysics: unresolved CBC, supernovae, and neutron stars

Search Methods

- CBC: matched filter
- Burst: identify coherent excess power in the data from multiple detectors
- CW: long time coherent matched filtering
- SGWB: cross-correlation

Packages

- Python (including libraries such as numpy, scipy, matplotlib)
- PyCBC
- lalsuite
- Healpix

Outlines

- Basics of Fourier Transformation
- Data analysis of SGWB
- Search of CBC

Basics of Fourier Transformation

Random processes

- A sequence of random variable
 - Instrument noise, SGWB
- Ensemble average $\langle x \rangle = \int x p(x) dx$ $p(x)$: probability density function
- Stationary random process: statistical properties do not change with time

$$\langle x \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} x(t) dt$$

Without loss of generality, we can set

$$\langle x \rangle = 0$$

Fourier Transformation

For a random processed data $x(t)$, Fourier transformation of $x(t)$ is defined as

$$\tilde{x}(f) = \int_{-\infty}^{\infty} x(t) e^{-i2\pi ft} dt$$

And the inverse Fourier transformation is defined as

$$x(t) = \int_{-\infty}^{\infty} \tilde{x}(f) e^{i2\pi ft} df$$

Sign convention:

Same as FFTW, Numpy, but different with numerical recipes

If $x(t)$ is real,

$$\tilde{x}(-f) = \tilde{x}^*(f)$$

Auto-correlation of x(t)

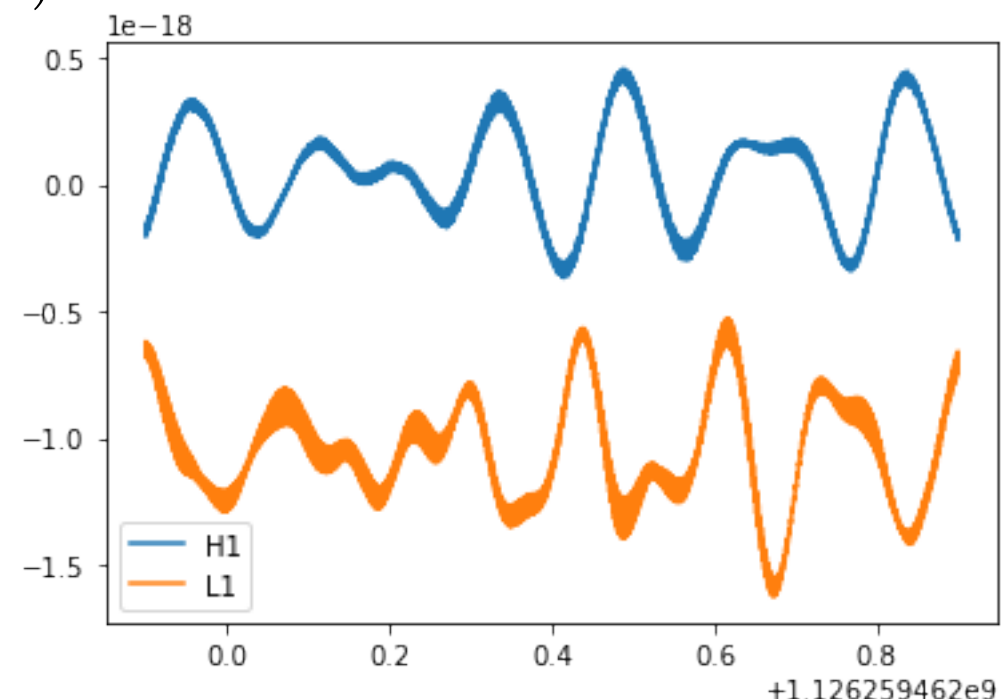
Auto-correlation of stationary x(t) is defined as

$$R(\tau) \equiv \langle x(t)x(t+\tau) \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} x(t')x(t'+\tau)dt' \quad \langle \rangle: \text{take average}$$

Stationary: the statistics depends only on time difference but not on the choice of time origin

$$\begin{aligned} R(\tau) &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} x(t')x(t'+\tau)dt' = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \int_{-\infty}^{\infty} \tilde{x}^*(f)\tilde{x}(f')e^{-i2\pi ft'}e^{i2\pi f'(t'+\tau)}dfdf'dt' \\ &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \tilde{x}^*(f)\tilde{x}(f')e^{i2\pi f'\tau}dfdf'\delta(f-f') \\ &= \int df \left[\lim_{T \rightarrow \infty} \frac{1}{T} \tilde{x}^*(f)\tilde{x}(f) \right] e^{i2\pi f\tau} \end{aligned}$$

White noise describes a random process whose mean is zero and whose autocorrelation is a delta-function



Power Spectrum Density

Define Power Spectrum Density

$$P(f) \equiv \lim_{T \rightarrow \infty} \frac{1}{T} |\tilde{x}(f)|^2$$

We have

$$R(\tau) = \int_{-\infty}^{\infty} df P(f) e^{i2\pi f\tau}$$

$$P(f) = \int_{-\infty}^{\infty} d\tau R(\tau) e^{-i2\pi f\tau}$$

Note:

1. The auto-correlation function is the FT of the power spectrum density

2. The auto-correlation function is real $R(\tau) = R(-\tau)$

3. $P(-f) = P(f)$

4. One-sided power spectrum density:

$$S(f) = P_{\text{one}}(f) = \begin{cases} 2P(f) & f \geq 0 \\ 0 & f < 0 \end{cases}$$

$$R(0) = \langle x^2 \rangle = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} x^2(t) dt = \int_0^{\infty} S(f) df$$

Stationary

$$\left\langle \tilde{x}^*(f') \tilde{x}(f) \right\rangle = \left\langle \int_{-\infty}^{\infty} x(t') e^{i2\pi f' t'} dt' \int_{-\infty}^{\infty} x(t) e^{-i2\pi f t} dt \right\rangle$$

Replace t by $t = t' + \tau$

$$\begin{aligned} \left\langle \tilde{x}^*(f') \tilde{x}(f) \right\rangle &= \left\langle \int_{-\infty}^{\infty} x(t') e^{i2\pi f' t'} dt' \int_{-\infty}^{\infty} x(t' + \tau) e^{-i2\pi f(t' + \tau)} d\tau \right\rangle \\ &= \int_{-\infty}^{\infty} e^{-i2\pi(f-f')t'} dt' \int_{-\infty}^{\infty} \langle x(t') x(t' + \tau) \rangle e^{-i2\pi f \tau} d\tau = P(f) \delta(f - f') \end{aligned}$$

Discrete Fourier Transform

The real data can not be infinite and continuous

$$\begin{array}{ccc} \tilde{x}(f) = \int_{-\infty}^{\infty} x(t) e^{-i2\pi ft} dt & \longrightarrow & \tilde{x}_k = \Delta t \sum_{j=0}^{N-1} x_j e^{-i2\pi t_j f_k} \quad x_j = x(t_j) \\ x(t) = \int_{-\infty}^{\infty} \tilde{x}(f) e^{i2\pi ft} df & & x_j = \Delta f \sum_{k=0}^{N-1} \tilde{x}_k e^{i2\pi t_j f_k} \quad \tilde{x}_k = \tilde{x}(f_k) \end{array}$$

We sample the time and frequency uniformly

$$t_j = j\Delta t = j \frac{T}{N}$$
$$f_k = k\Delta f = k \frac{1}{T}$$

Then

$$t_j f_k = j \frac{T}{N} k \frac{1}{T} = \frac{jk}{N}$$

Discrete Fourier Transform

$$\tilde{x}_k = \Delta t \sum_{j=0}^{N-1} x_j e^{-i2\pi jk/N}$$
$$x_j = \Delta f \sum_{k=0}^{N-1} \tilde{x}_k e^{i2\pi jk/N}$$



What numerical libraries do for us

Discrete Fourier Transform

Noise power spectrum

$$\left\langle \tilde{n}^*(f') \tilde{n}(f) \right\rangle = \frac{1}{2} \delta(f - f') S_n(|f|)$$

In DFT

$$\left\langle \tilde{n}_i^* \tilde{n}_j \right\rangle = \frac{1}{2} D_{ij} S_n(|f|)$$

Normalization:

$$\int \delta(f - f') df = 1 \rightarrow \sum_i D_{ii} \Delta f = 1$$

Thus, we have

$$D_{ij} = 0 \quad i \neq j$$

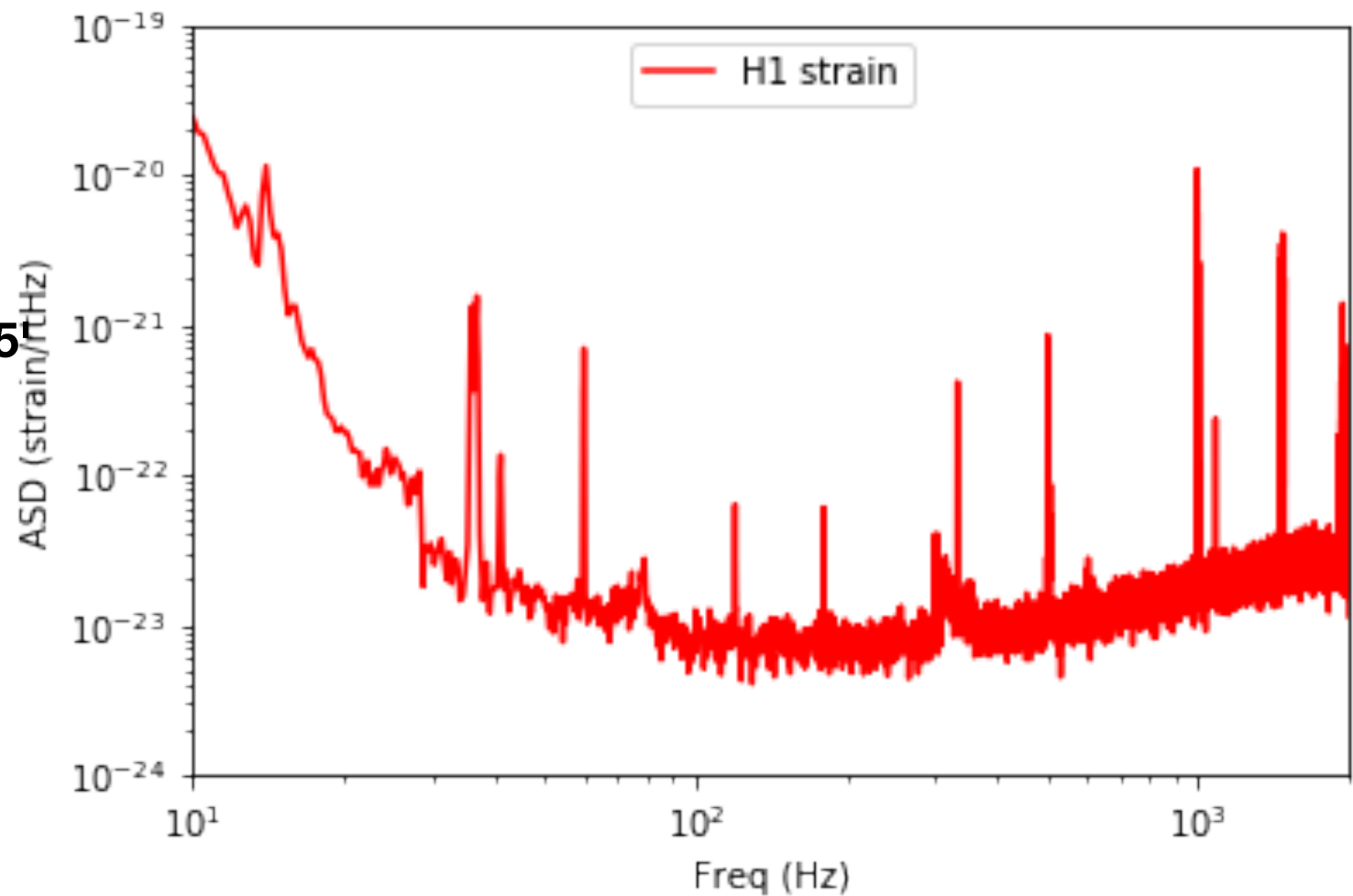
$$D_{ij} = \frac{1}{\Delta f} = T \quad i = j \quad \left\langle \tilde{n}_i^* \tilde{n}_j \right\rangle = \frac{1}{2\Delta f} S_n(|f_i|) \delta_{ij} = \frac{T}{2} S_n(|f_i|) \delta_{ij}$$

Variance

$$\sigma_i^2 = \sigma^2(f_i) = \left\langle |n_i^R|^2 \right\rangle = \left\langle |n_i^I|^2 \right\rangle = \frac{1}{4\Delta f} S_n(f_i)$$

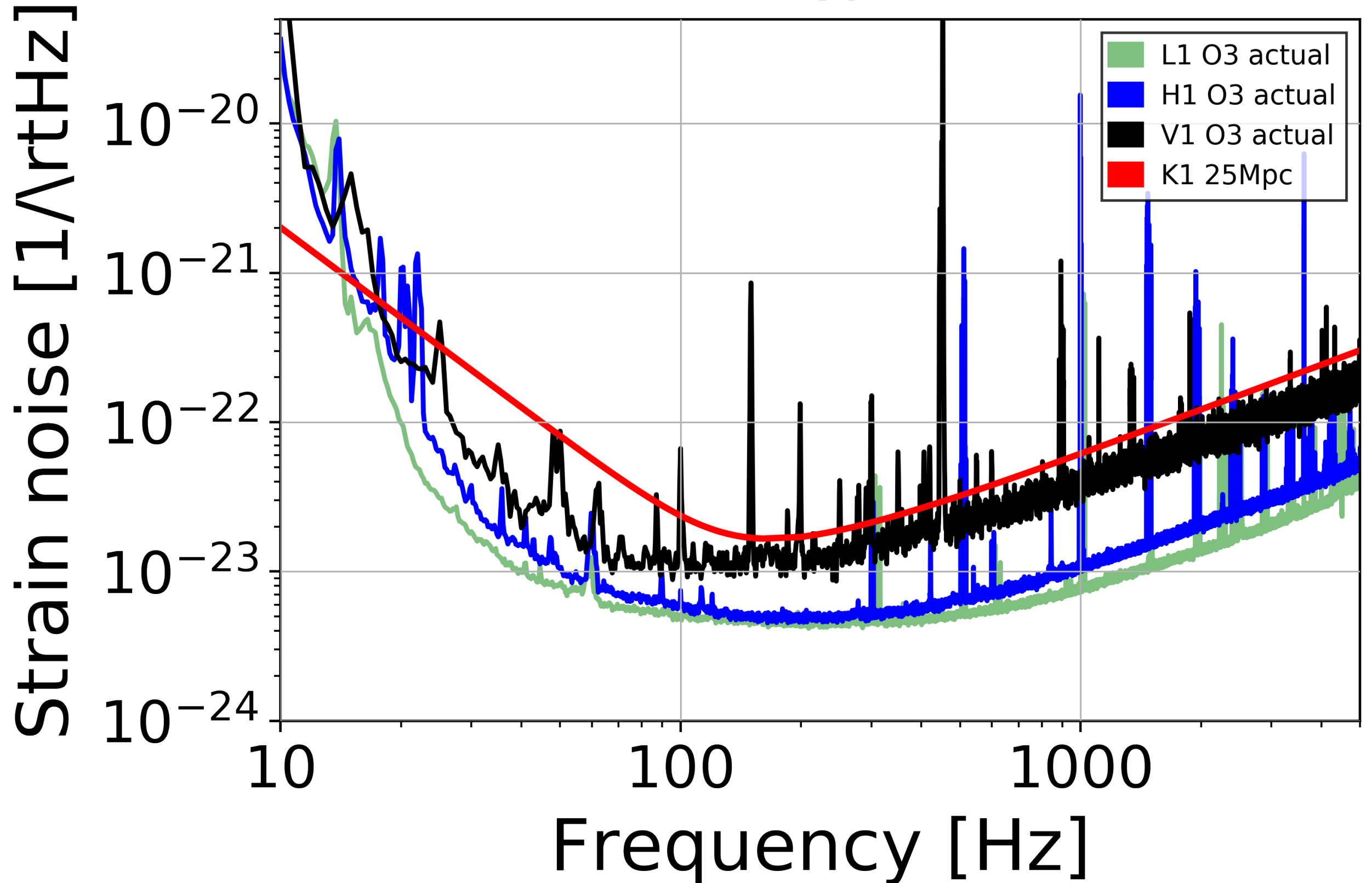
Sensitivity Curve

```
import numpy as np
import readligo as rl
import matplotlib.pyplot as plt
import h5py
import matplotlib.mlab as mlab
fn_H1 = 'H-H1_LOSC_4_V1-1126259446-32.hdf5'
strain_H1, time_H1, chan_dict_H1 =
rl.loaddata(fn_H1, 'H1')
fs = 4096
NFFT = 4*fs
noverlap = 0
PSD_H1, freqs = mlab.psd(strain_H1, Fs = fs,
NFFT = NFFT, noverlap=noverlap)
```



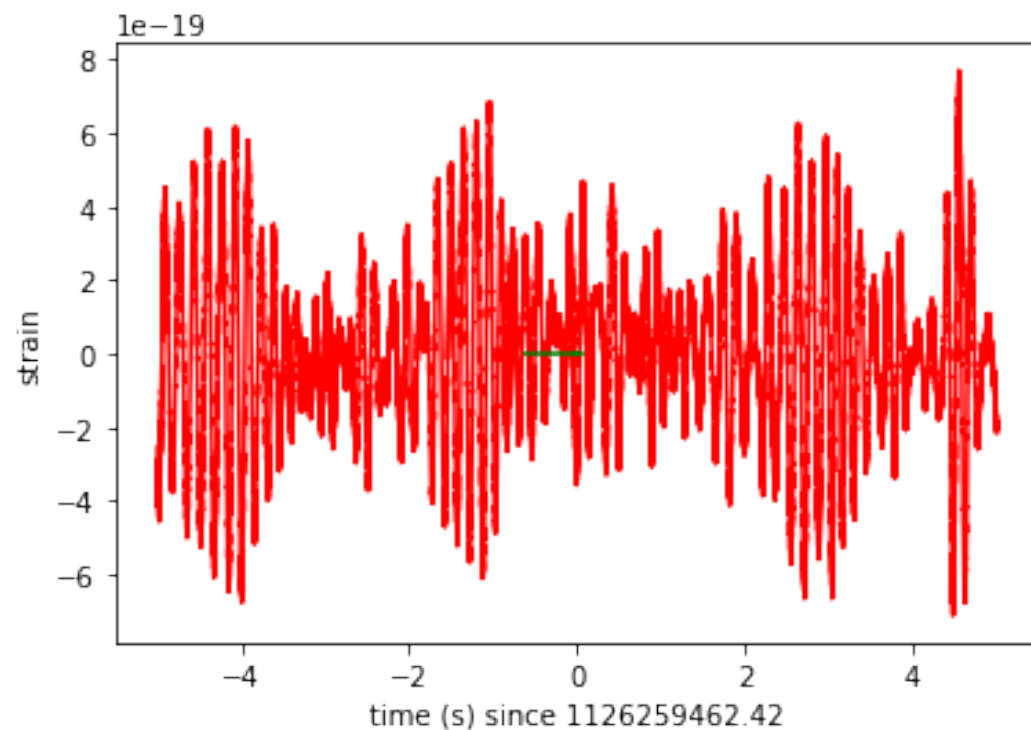
O3 Sensitivity Curves

LIGO

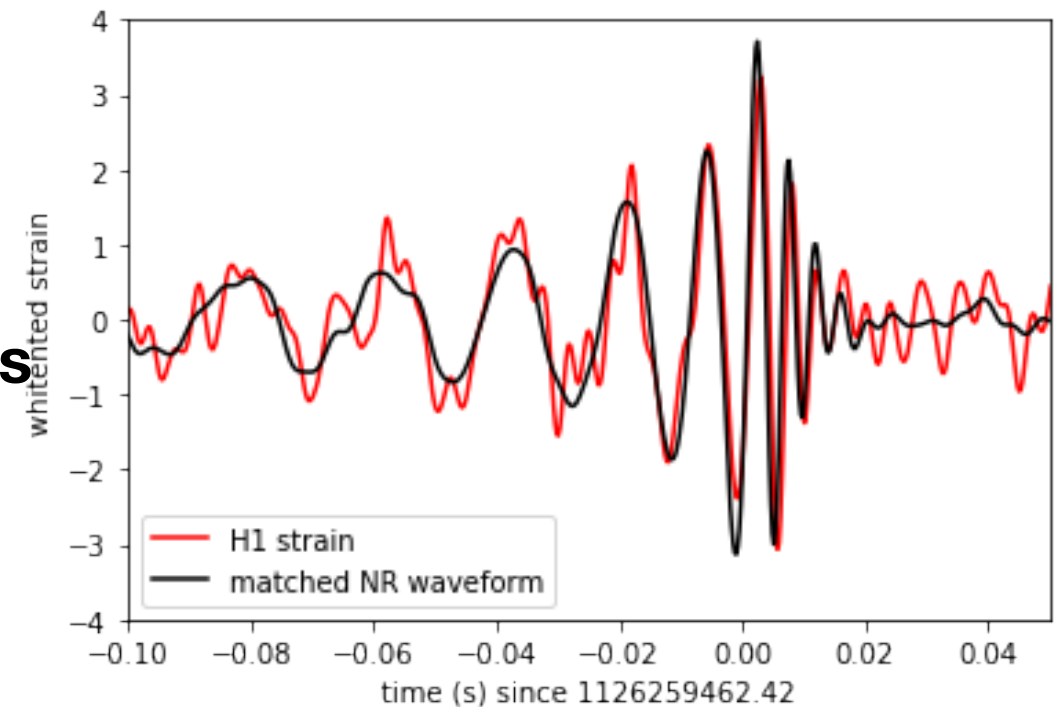


Stochastic Signals

Stochastic signal is randomly produced by a lot of number of weak, unresolved sources



After process



What if magnitude of signal is much smaller than this?

Search of SGWB

Cross-Correlation of data in two detectors

- Early papers of using cross-correlation to search SGWB: Grishchuk 1976; Helling and Downs 1983
- Recent papers:
 - Mitra et al., Gravitational wave radiometry: Mapping a stochastic gravitational wave background (2008)
 - Cornish & Romano, Towards a unified treatment of gravitational-wave data analysis (2013)
 - Thrane et al. Probing the anisotropies of a stochastic gravitational-wave background using a network of ground-based laser interferometers (2009)
 - Ain, Suresh & Mitra, Very fast stochastic gravitational wave background map-making using folded data(2018)

Cross-Correlation of data in two detectors

Consider we have two detectors:

$$d_1 = h_1 + n_1$$

$$d_2 = h_2 + n_2$$

Expected value of the cross-correlation

$$\langle C_{12} \rangle = \langle d_1 d_2 \rangle = \langle h_1 h_2 \rangle + \langle h_1 n_2 \rangle + \langle n_1 h_2 \rangle + \langle n_1 n_2 \rangle = \langle h_1 h_2 \rangle$$

GW Radiation

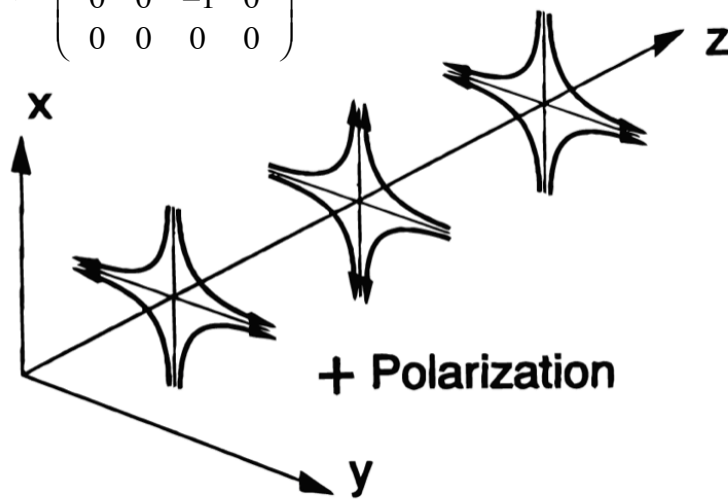
Metric perturbation by GW can be written as superposition of plan waves

$$h_{ab}(t, \vec{x}) = \int_{-\infty}^{\infty} df \int_{S^2} d^2 \hat{n} h_{ab}(f, \hat{n}) e^{i2\pi f(t + \hat{n} \cdot \vec{x}/c)}$$

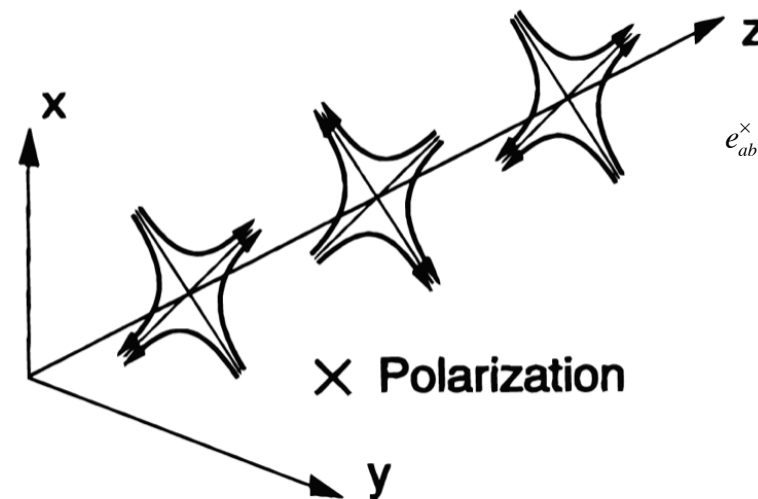
Expand the GW in the standard + and x polarization

$$h_{ab}(f, \hat{n}) = h_+ e_{ab}^+(\hat{n}) + h_{\times} e_{ab}^{\times}(\hat{n})$$

$$e_{ab}^+(\hat{n}) = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$



$$e_{ab}^{\times}(\hat{n}) = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$



Abramovici et al. 1992

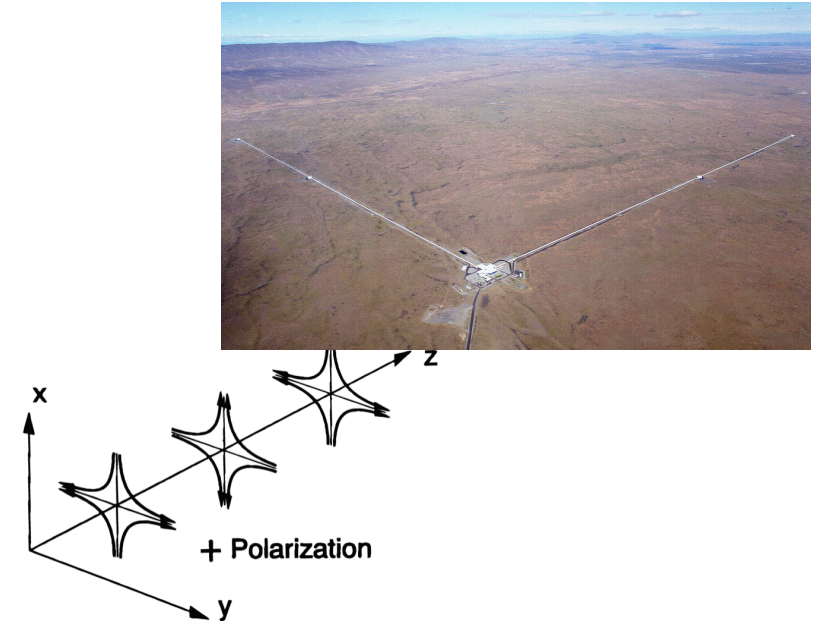
Detector Response

Detector Tensor

$$d^{ab} \equiv \frac{1}{2} \begin{pmatrix} \hat{x}^a \hat{x}^b - \hat{y}^a \hat{y}^b \end{pmatrix}$$

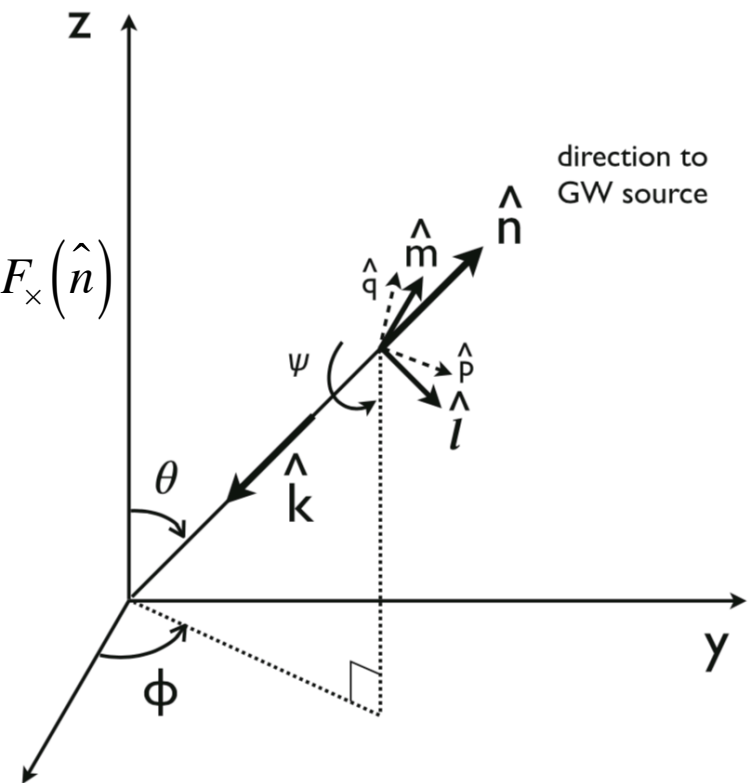
$$e_{ab}^+(\hat{n}) = \hat{l}_a \hat{l}_b - \hat{m}_a \hat{m}_b$$

$$e_{ab}^\times(\hat{n}) = \hat{l}_a \hat{m}_b + \hat{m}_a \hat{l}_b$$



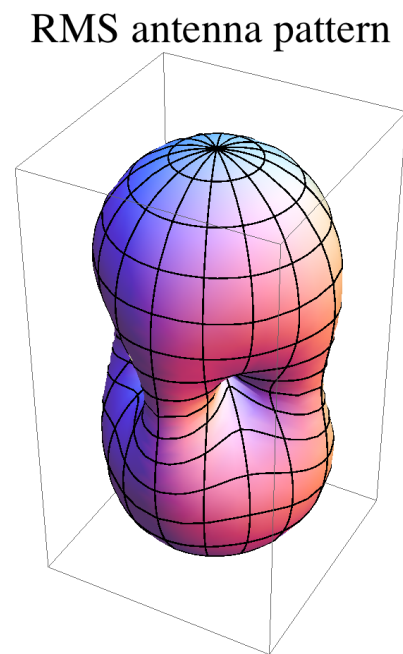
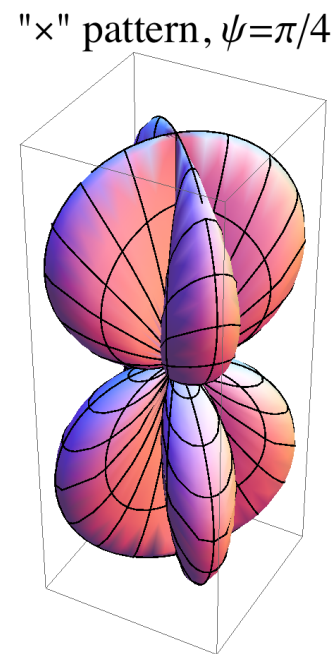
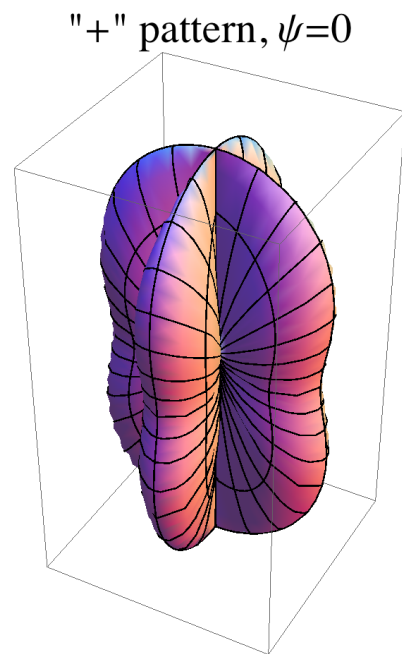
Response of interferometer

$$h(f, \hat{n}) = \frac{\Delta L}{L} = d^{ab} h_{ab}(f, \hat{n}) = h_+ d^{ab} e_{ab}^+(\hat{n}) + h_\times d^{ab} e_{ab}^\times(\hat{n}) = h_+ F_+(\hat{n}) + h_\times F_\times(\hat{n})$$



Antenna Pattern

$F_{+/\times}$: antenna pattern



L. Singer arXiv:1501.03765

```
import pycbc.detector as Detector  
f_plus, f_cross = Detector.overhead_antenna_pattern(ra, dec, polarization)
```


Cross-Correlation

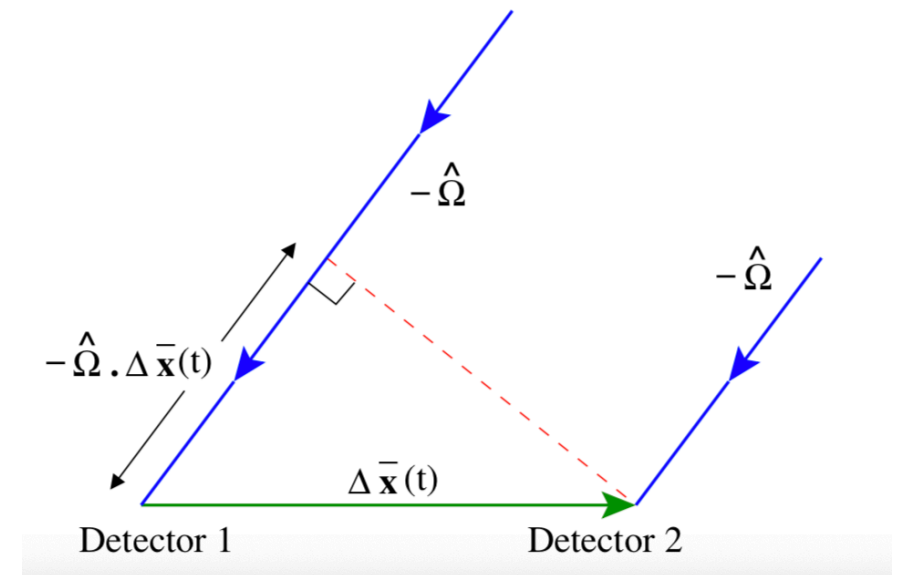
Cross-Correlation of data in two detectors is very similar to the technique in radio astronomy. For a given source in the sky, the GW signal will arrive the two detector with a time delay.

$$\Delta t = \frac{\hat{\Omega} \cdot \Delta \mathbf{x}}{c} \quad \Delta \mathbf{x} : \text{baseline}$$

The resolution of the image

$$\Delta \theta \sim \frac{\lambda_{GW}}{\Delta x \sin \theta}$$

The baseline changes as the Earth rotates. It changes the delay and cross-correlation of two GW signal as well.



Mitra et al. 2008

Short Time Fourier Transform

Time series data of two detectors

$$d_I(t) = h_I(t) + n_I(t) \quad I = 1, 2$$

Short time Fourier transform

$$\tilde{d}(f; T) = \int_{T-\tau/2}^{T+\tau/2} d(t) e^{-i2\pi ft} dt$$

τ is the duration of the time segment. It should be much longer than the traveling time of GW between two detectors but smaller enough so that the change of antenna-pattern function is not significant.

Now we define the cross-correlation of the output as

$$C(f; T) \equiv \tilde{d}_1^*(f, T) \tilde{d}_2(f, T)$$

The expected value

$$\langle C(f; T) \rangle = \frac{\tau}{2} \langle \tilde{h}_1^*(f, T) \tilde{h}_2(f, T) \rangle$$

Inverse FFT of cross-correlation give the delay information

$$\begin{aligned}
 C(\tau) &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} h_1(t') h_2(t' + \tau) dt' = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} \int_{-\infty}^{\infty} \tilde{h}_1^*(f) \tilde{h}_2(f') e^{-i2\pi f t'} e^{i2\pi f'(t' + \tau)} df df' dt' \\
 &= \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \tilde{h}_1^*(f) \tilde{h}_2(f') e^{i2\pi f' \tau} df df' \delta(f - f') \\
 &= \int df \left[\lim_{T \rightarrow \infty} \frac{1}{T} \tilde{h}_1^*(f) \tilde{h}_2(f) \right] e^{i2\pi f \tau}
 \end{aligned}$$

$$\begin{array}{ccc}
 C(\tau) & \xrightleftharpoons[\text{iFT}]{\text{FT}} & \langle \tilde{h}_1^* \tilde{h}_2 \rangle
 \end{array}$$

Statistic of Sources of SGWB

GW radiation from different directions of the sky are uncorrelated, unpolarized, Gaussian-stationary

$$\left\langle h_A(f, \hat{n}) \right\rangle = 0 \quad \mathbf{A: + or x}$$

$$\left\langle h_A^*(f, \hat{n}) h_{A'}(f', \hat{n}') \right\rangle = \frac{1}{4} \mathbf{P}(f, \hat{n}) \delta(f - f') \delta_{AA'} \delta(\hat{n}, \hat{n}')$$

Define

$$S_h(f) = \int d^2 \hat{n} \mathbf{P}(f, \hat{n})$$

and

$$\Omega_{gw}(f) \equiv \frac{f}{\rho_c} \frac{d\rho_{gw}}{df} = \frac{2\pi^2}{3H_0^2} f^3 S_h(f) \quad \rho_c = \frac{3c^2 H_0^2}{8\pi G} : \text{critical energy density of the universe}$$

We can assume power from all the directions of the sky have the same frequency response

$$\mathbf{P}(f, \hat{n}) = H(f) P(\hat{n})$$

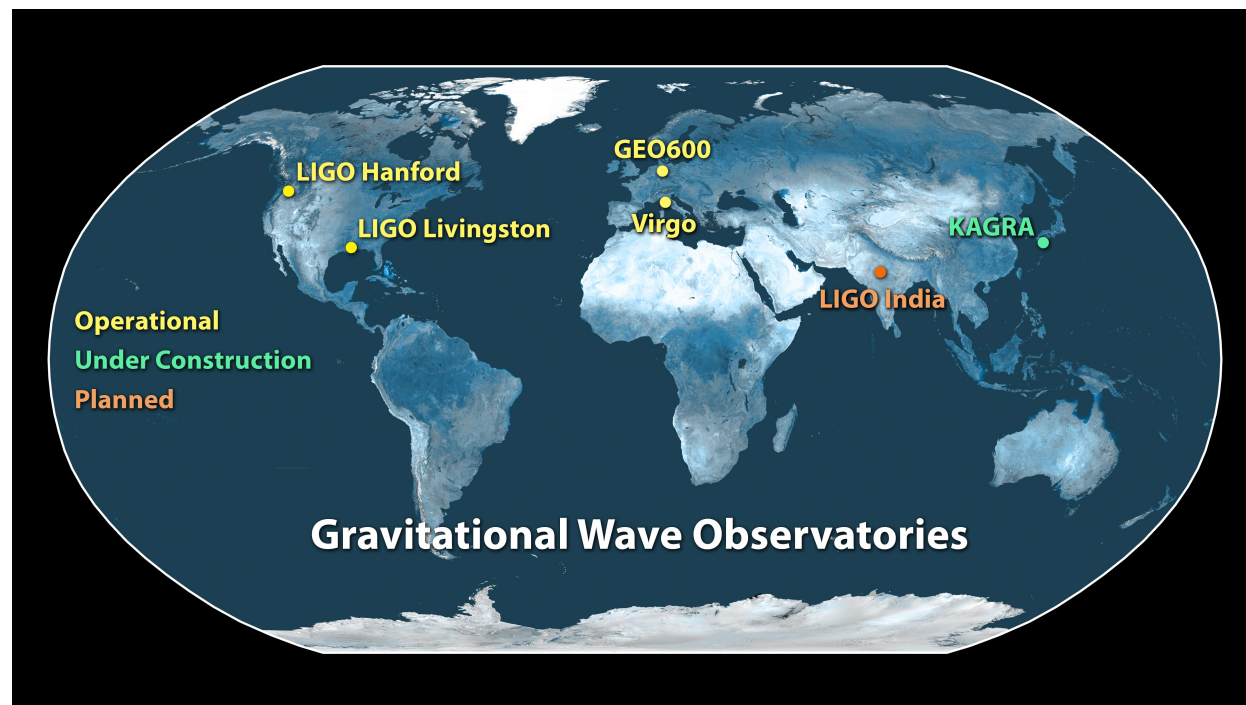
Directional Search of SGWB

- Mechanisms of GW production
- Astrophysical distribution of GW sources in galaxies
- Galaxy formation and distribution of the large scale structures of the universe

Cross-Correlation of signals

$$\langle h_1^*(f;T) h_2(f;T) \rangle = \int d^2 \hat{n} \sum_A F_{1A}(\hat{n};T) F_{2A}(\hat{n};T) \mathbf{P}_A(\hat{n},f) e^{i2\pi f(\hat{n} \cdot \Delta \vec{x}/c)}$$

Correlation of signal from two detectors



<https://www.ligo.caltech.edu/image/ligo20160211c>



Visibility $V(\Delta \vec{x}) = \int d\hat{\Omega} \Delta I(\hat{\Omega}) A(\hat{\Omega}) e^{i2\pi f \Delta \vec{x} \cdot \hat{\Omega}/c}$

$$\gamma(f, t, \hat{n}) = \sum_A F_{1A}(\hat{n}; T) F_{2A}(\hat{n}; T) e^{i2\pi f(\hat{n} \cdot \Delta \vec{x}/c)}$$

```

1 import numpy as np
2 import healpy as hp
3 nside = 32
4 npix=hp.nside2npix(nside)
5 ipix=np.arange(npix)
6 theta, phi=hp.pix2ang(nside, ipix)
7 ra = -phi
8 dec = np.pi/2.0 - theta

```

```

1 from pycbc.detector import Detector
2 from lal.antenna import AntennaResponse
3 from pycbc.detector import Detector
4
5
6 def gamma_map(freq, nside, detector1, detector2, polarization, gps_time, ra, dec):
7     npix=len(ra)
8     omega= 4*np.pi/npix
9     antenna_p=np.zeros([len(ra)], dtype=complex)
10    antenna_c=np.zeros([len(ra)], dtype=complex)
11
12    det1=Detector(detector1)
13    det2=Detector(detector2)
14    timedelay=det1.time_delay_from_detector(det2, ra, dec, gps_time)
15
16    for i in range(npix):
17        phase = 2.0 * np.pi * freq * timedelay[i]
18        resp1 = AntennaResponse(detector1, ra[i], dec[i], psi=polarization, scalar=True, vector=True, times=gps_time)
19        resp2 = AntennaResponse(detector2, ra[i], dec[i], psi=polarization, scalar=True, vector=True, times=gps_time)
20        antenna_p[i] = (resp1.plus*resp2.plus)*complex(np.cos(phase), np.sin(phase)) #*2.5
21        antenna_c[i] = (resp1.cross*resp2.cross)*complex(np.cos(phase), np.sin(phase)) #*2.5
22    return antenna_p, antenna_c

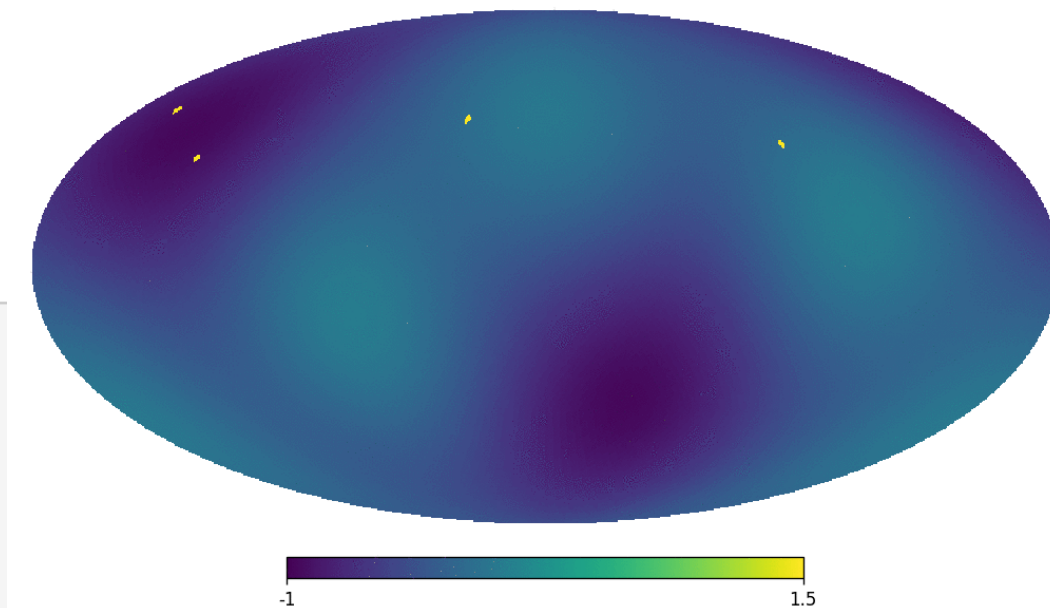
```

```

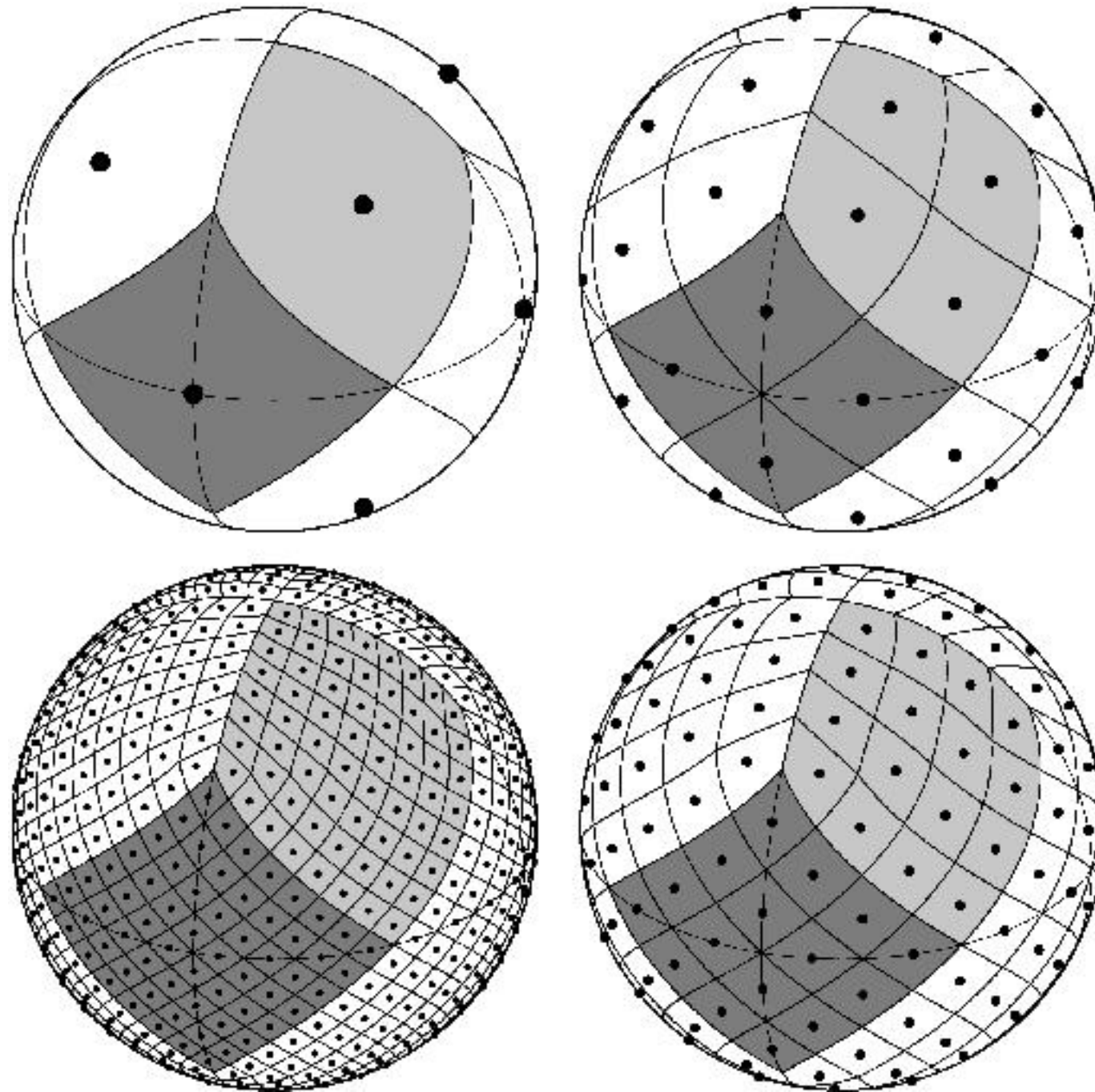
18 gmst = pycbc.detector.gmst_accurate(gps)

```

LH f=0

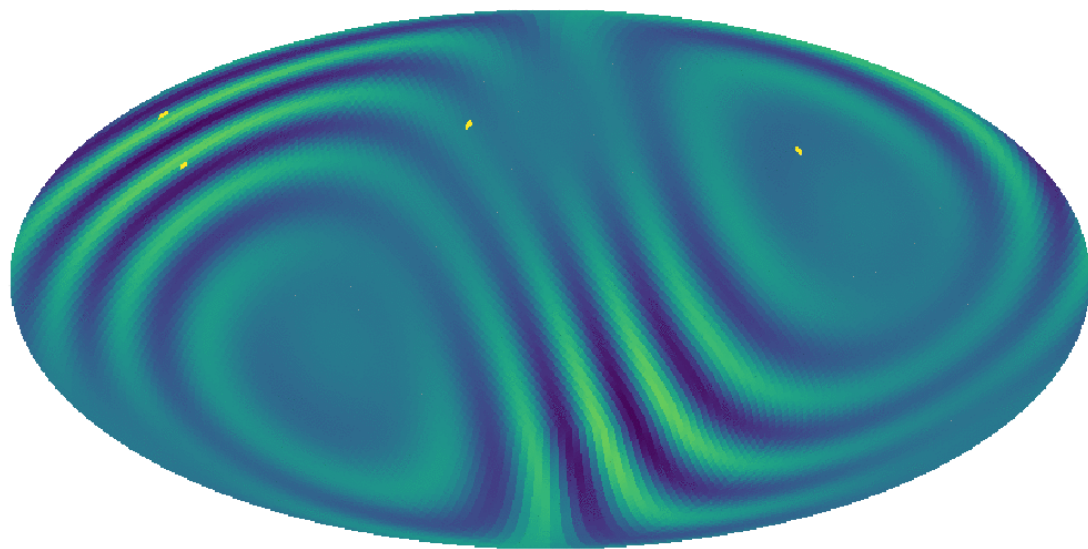


Healpix

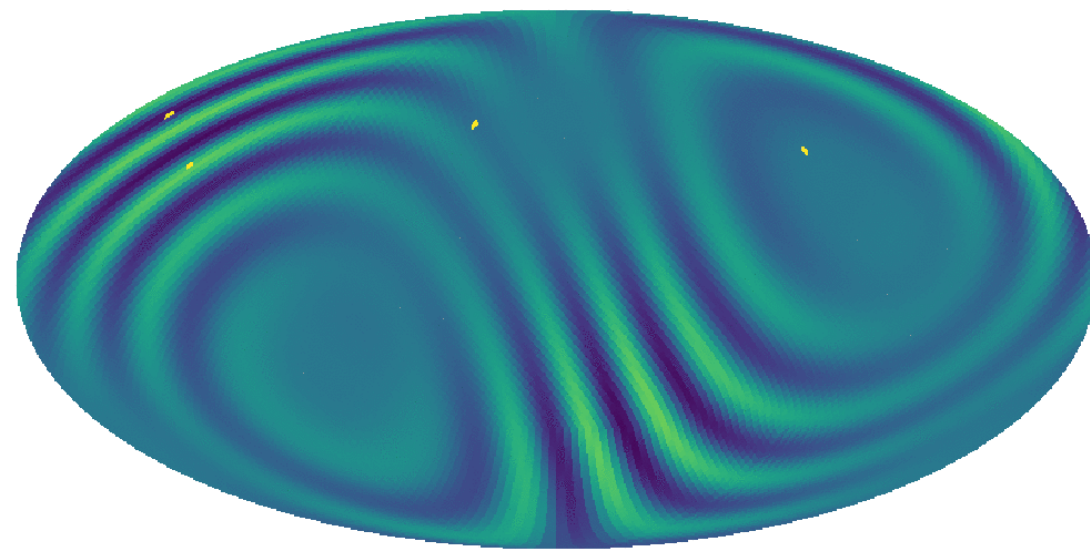


$$\gamma(f, t, \hat{n}) = \sum_A F_{1A}(\hat{n}; T) F_{2A}(\hat{n}; T) e^{i2\pi f(\hat{n} \cdot \Delta \vec{x}/c)}$$

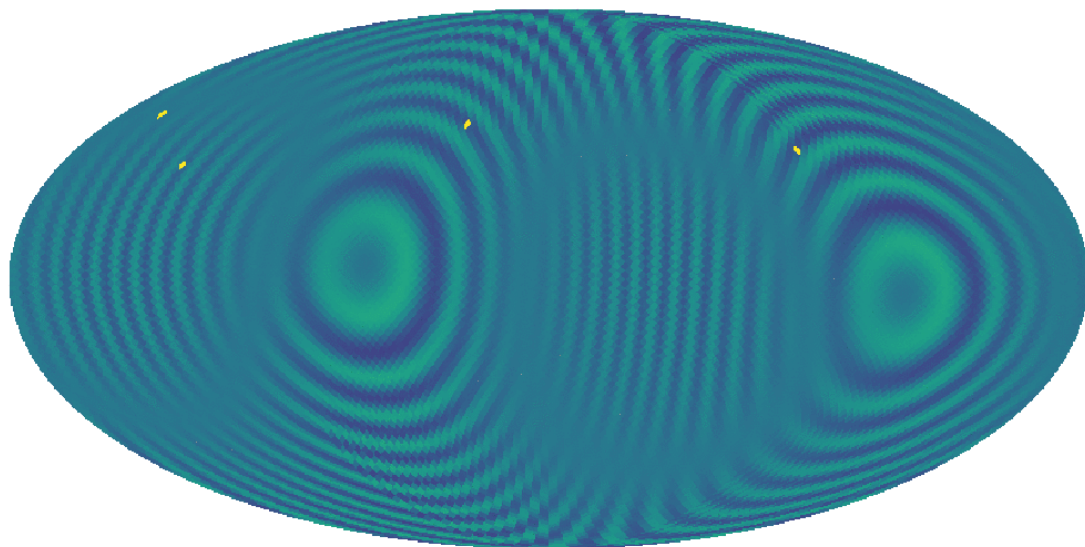
LH real f=200



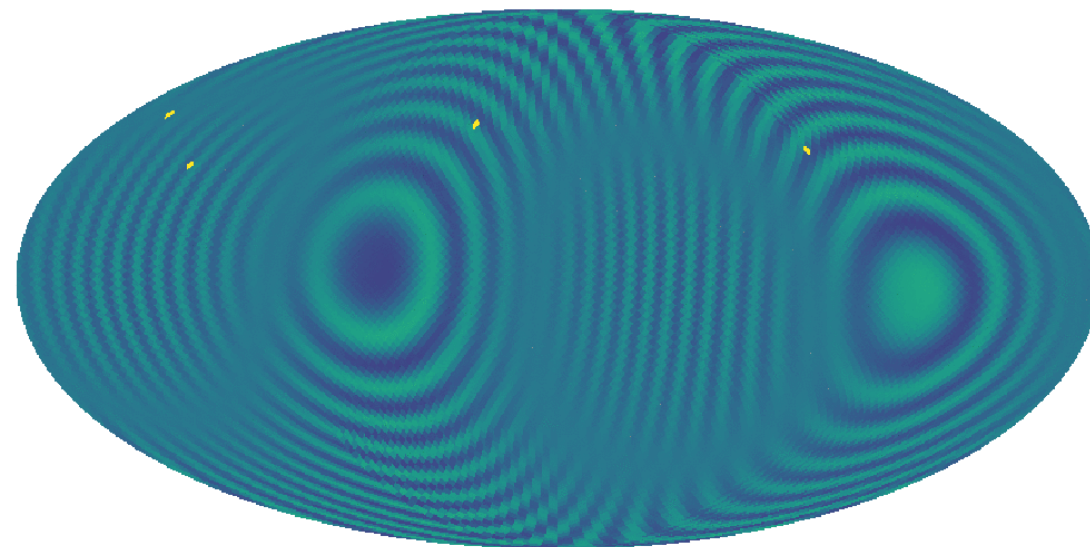
LH Imag f=200



VK real f=200



VK Imag f=200



Dirty Map

$$C = KP + n$$

Correlation **Map** Noise

Dirty Map :

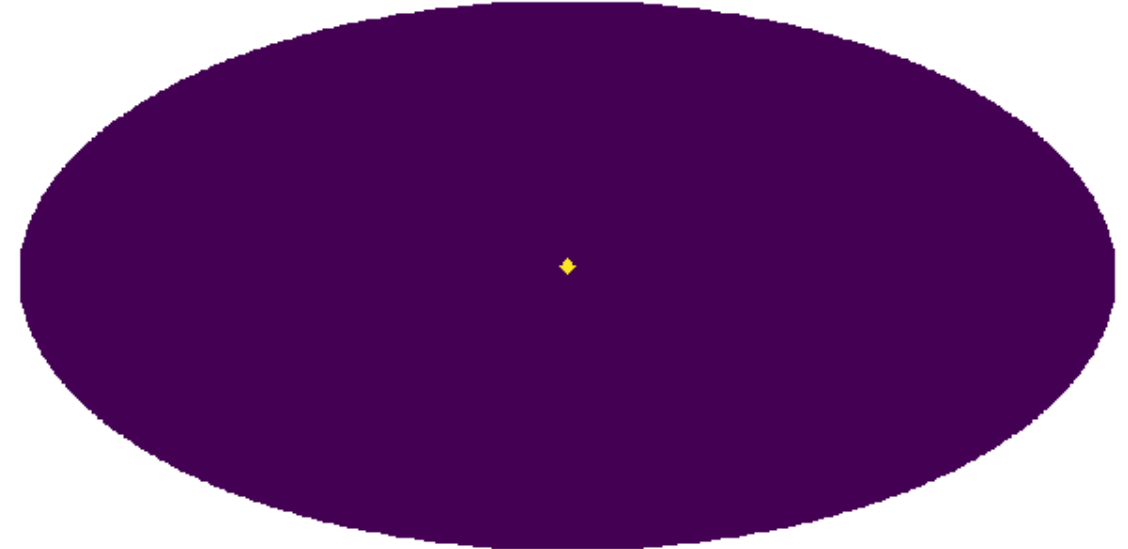
$$P_{\text{Dirty}} = K^{\dagger} N^{-1} C$$

Covariant matrix of noise

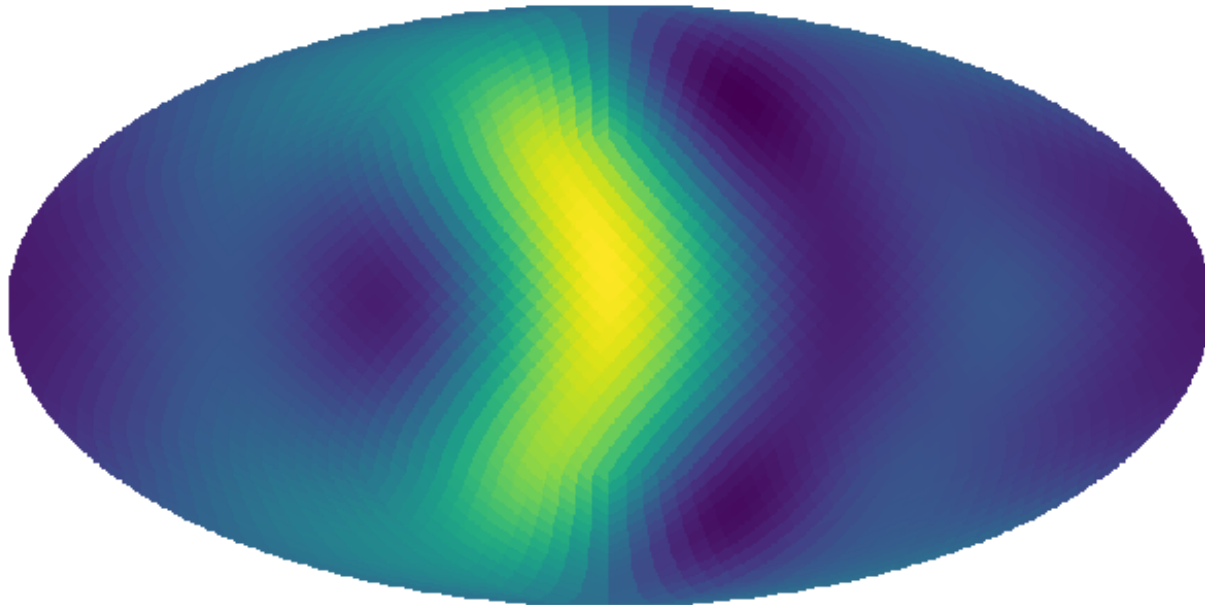
$$N_{fT, f'T'} = \langle C_{fT}^* C_{f'T'} \rangle - \langle C_{fT} \rangle \langle C_{f'T'} \rangle \sim \frac{\tau^2}{4} \delta_{ff'} \delta_{TT'} S_1(f; T) S_2(f; T)$$

Dirty Map

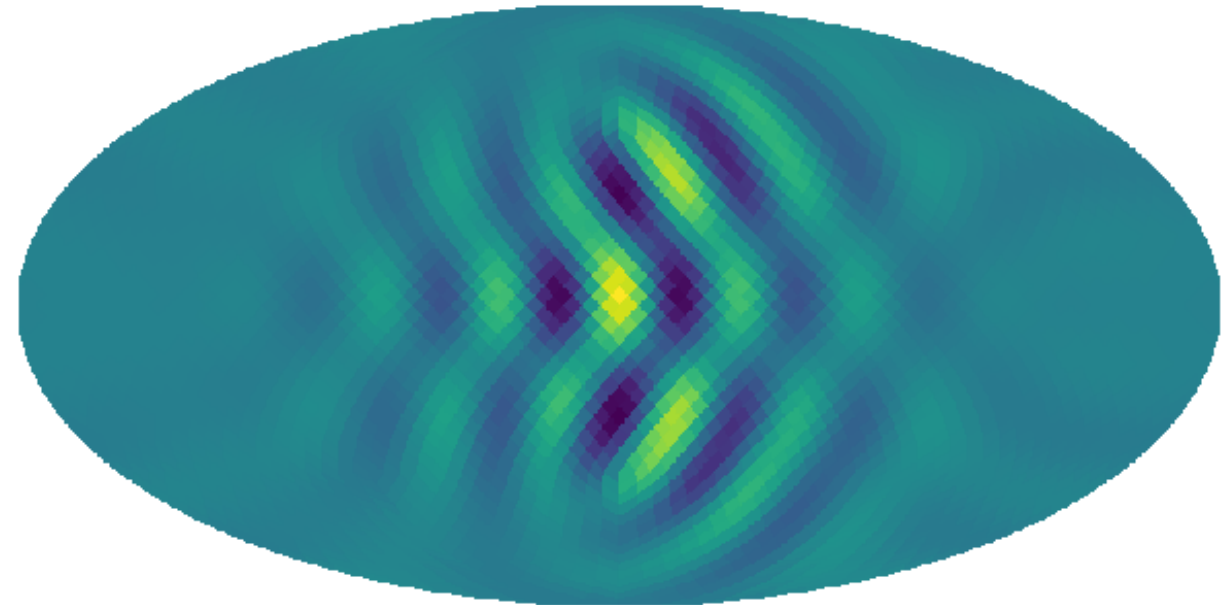
injection



L-H, $f=40\text{Hz}$

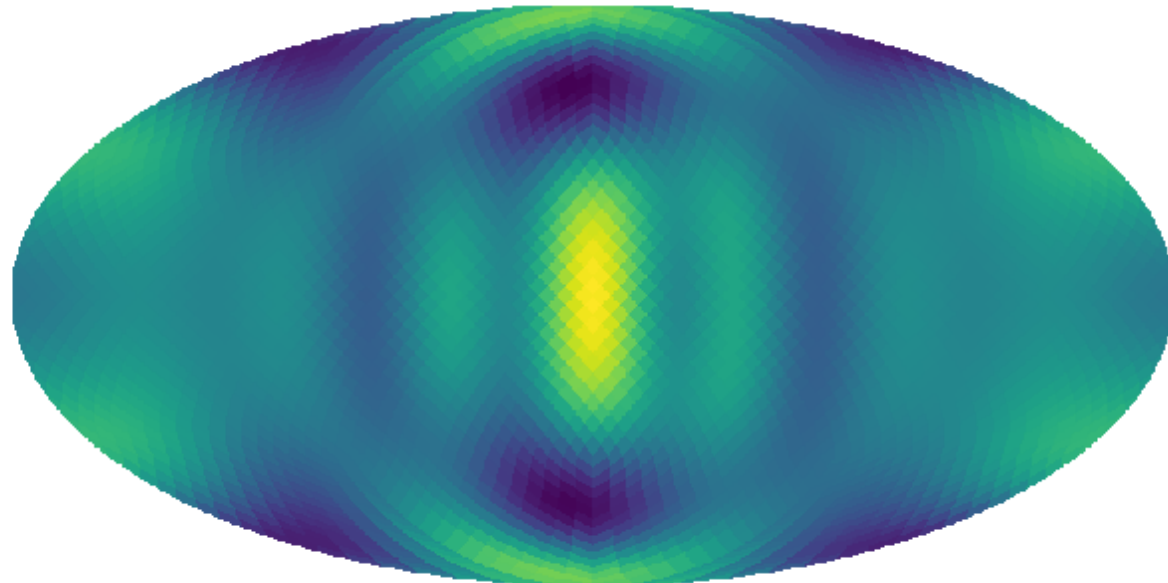


L-H, $f=200\text{Hz}$

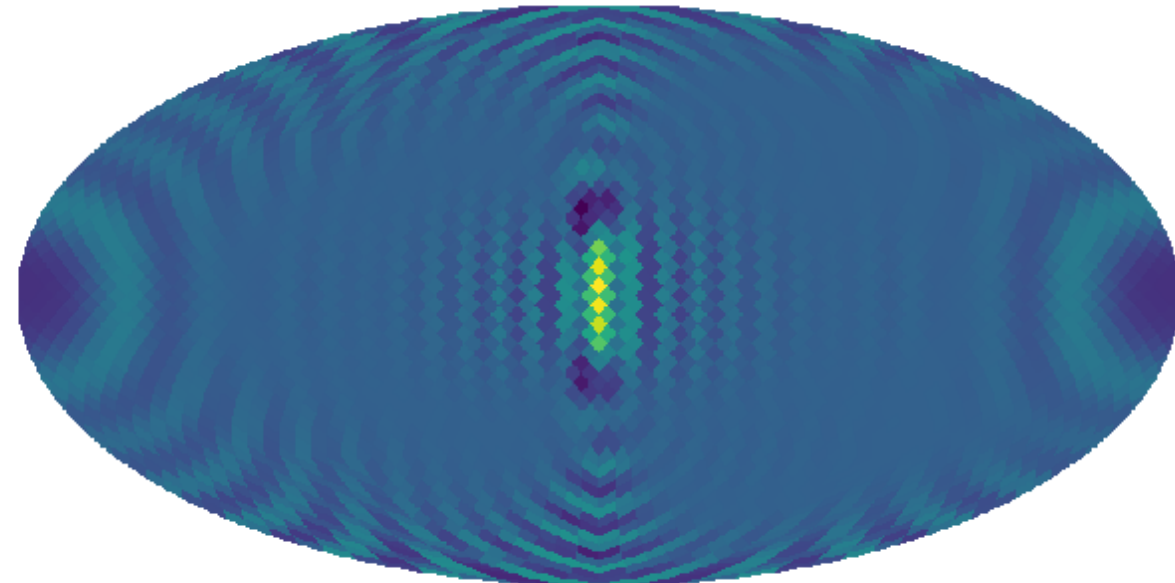


Dirty Map

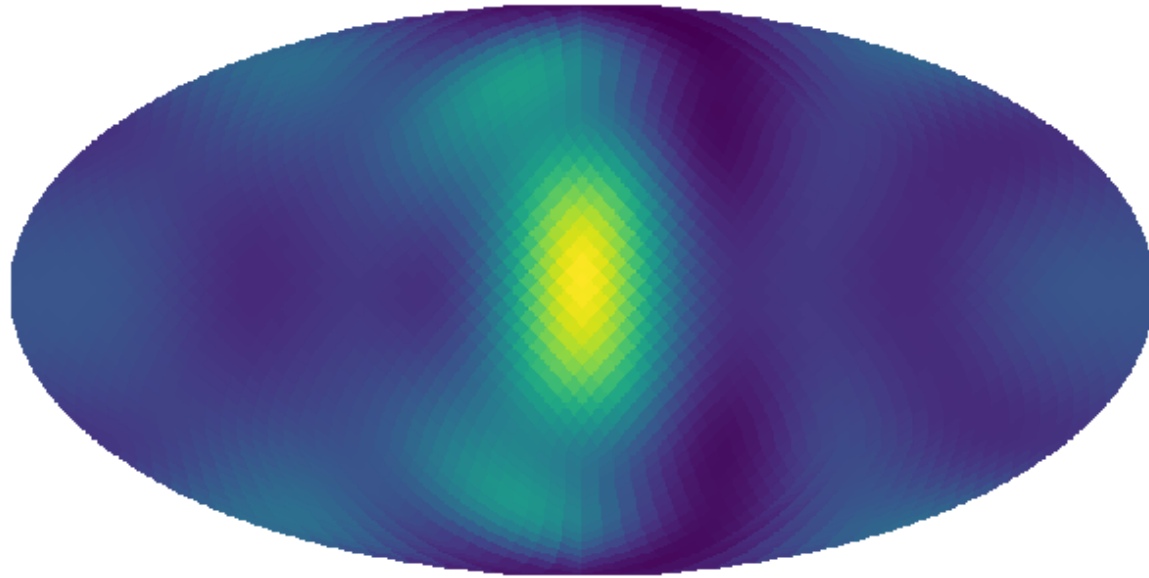
V-K, $f=40\text{Hz}$



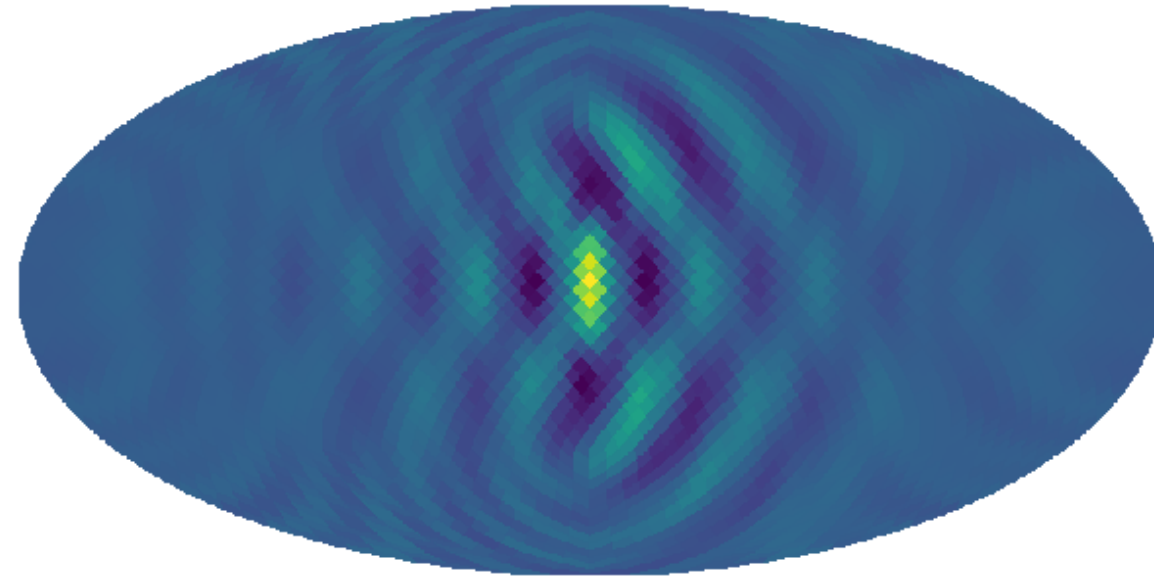
V-K, $f=200\text{Hz}$



L-H-V-K, $f=40\text{Hz}$



L-H-V-K, $f=200\text{Hz}$



Clean Map

To estimate the map P , we can minimize the chi square

$$\chi^2(P) = \left(C_{fT}^* - \langle C_{fT}^*(P) \rangle \right) N_{fT, f'T'}^{-1} \left(C_{fT} - \langle C_{fT}(P) \rangle \right)$$

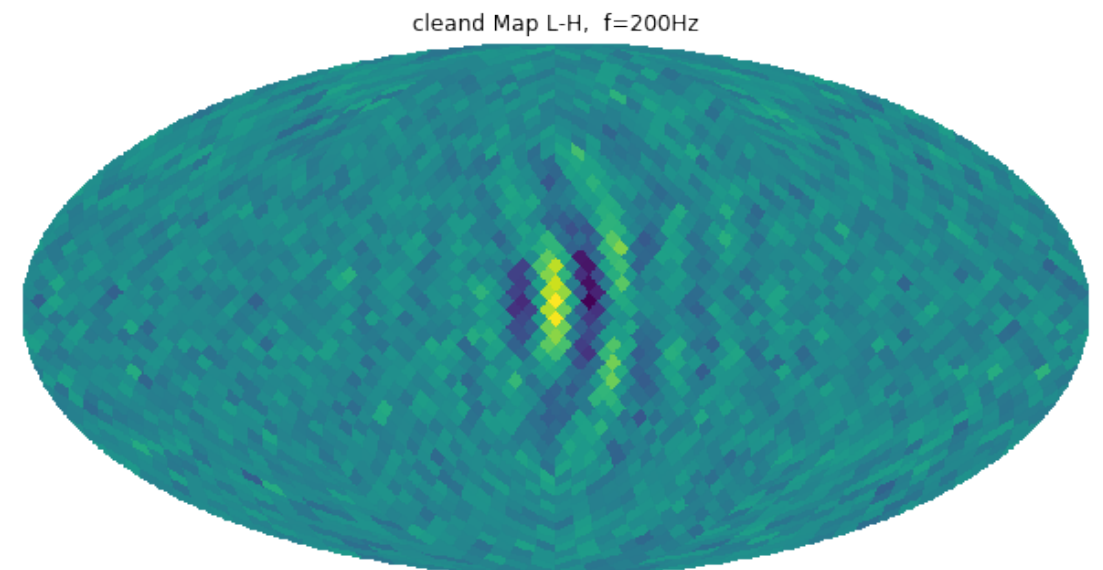
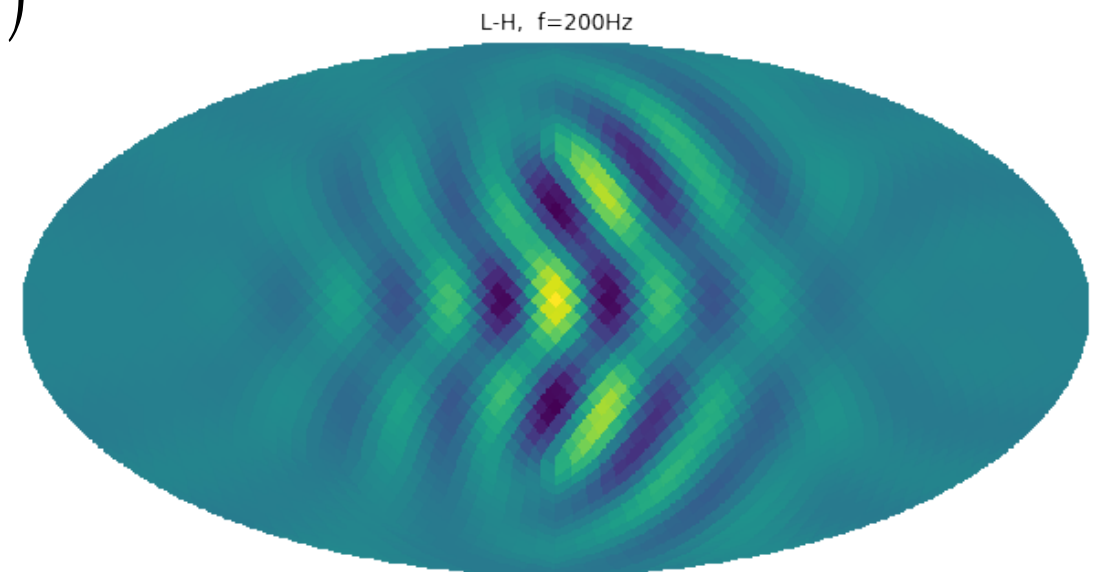
Clean map

$$\hat{P} = \Gamma^{-1} X$$

$$\Gamma = 4 \sum_{fT} \frac{H^2(f)}{P_1(f, T) P(f, T)} \gamma_p^*(f, T) \gamma_{p'}(f, T)$$

Or in matrix

$$\Gamma = \left(K^\dagger N^{-1} K \right)$$



Clean Map

Indeed, Γ is a singular matrix. We need SVD method to inverse the matrix.

$$\Gamma = USU^\dagger$$

$$S = \text{diag}(s_i) \quad \text{Si are the eigenvalues of } \Gamma$$

Set a threshold $s_{\min}$, replace $s_i < s_{\min}$ by $s_{\min}$.

$$\Gamma' = US'U^\dagger$$