

China Winter School on LISA MLDC – Lecture 2

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Plan

- Matched filtering
 - Special case: sinusoid signal (amplitude and initial phase parameters)
- Matched filtering in Fourier domain
- Lab exercises

Least-squares fitting

- Given data: $(x_i, y_i); i = 0, 1, \dots, N - 1$
 - $\bar{y} = (y_0, y_1, \dots, y_{N-1})$
- Assumption: $y_i = f(x_i) + n_i$
 - n_i : Noise (random variable)
 - n_i, n_j : statistically independent if $i \neq j$
 - (Mean) $E[n_i] = 0$
 - (Variance) $E[n_i^2] = \sigma^2$
- Least squares fitting: $\min_{\Theta} \|\bar{y} - \bar{s}(\Theta)\|^2$
 - $\|\bar{x}\|^2 = \sum_{i=0}^{N-1} x_i^2$
 - Example: $\bar{s}(\Theta) = a\bar{x} + b$
- $\|\bar{y} - \bar{s}(\Theta)\|^2 = \|\bar{y}\|^2 - 2\langle \bar{y}, \bar{s}(\Theta) \rangle + \|\bar{s}(\Theta)\|^2$

Matched filtering

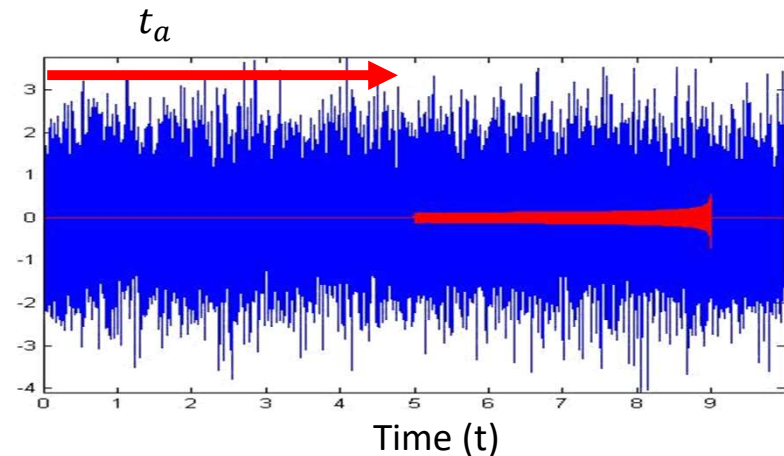
- Independent variable x is time t

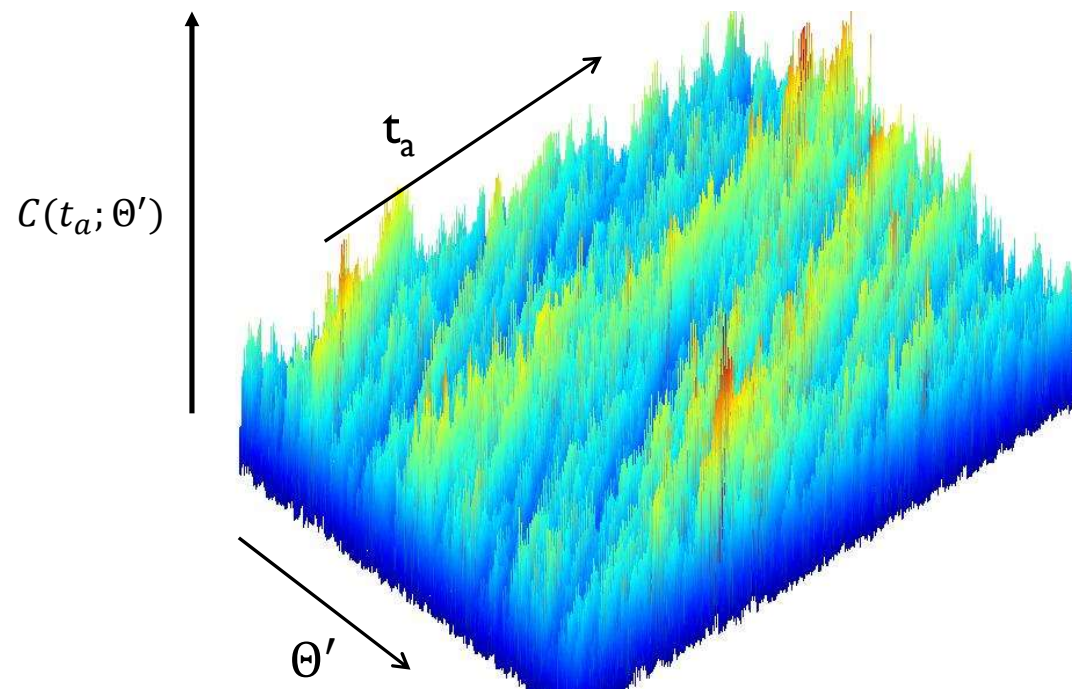
- $y(t) = s(t; \Theta) + n(t)$
- $y_i = y(t_i)$

- $\Theta = \{t_a, \Theta'\}; s(t; \Theta) = s(t - t_a; \Theta')$

- Least-squares fitting over the time of arrival parameter is a filtering operation

- $\langle \bar{y}, \bar{s} \rangle = \left(\frac{1}{\delta t} \right) \int_0^T dt y(t) s(t - t_a; \Theta') = C(t_a; \Theta') : \text{Correlation of data with signal}$
- For fixed Θ' , $\min_{t_a} \|\bar{y} - \bar{s}\|^2$ is equivalent to $\max_{t_a} \langle \bar{y}, \bar{s} \rangle$
- Convolution theorem: $F[C(t)](f) = F[y(t)](f) \times F[s(t; \Theta')](f)$
- $F[a(t)](f) = \int_{-\infty}^{\infty} dt a(t) e^{-2\pi i f t}$ (Fourier transform)
- Caution: very loose notation! Assumes $T \gg \text{signal length}$





Finding the global maximum of $C(t_a; \Theta')$ is a difficult optimization problem

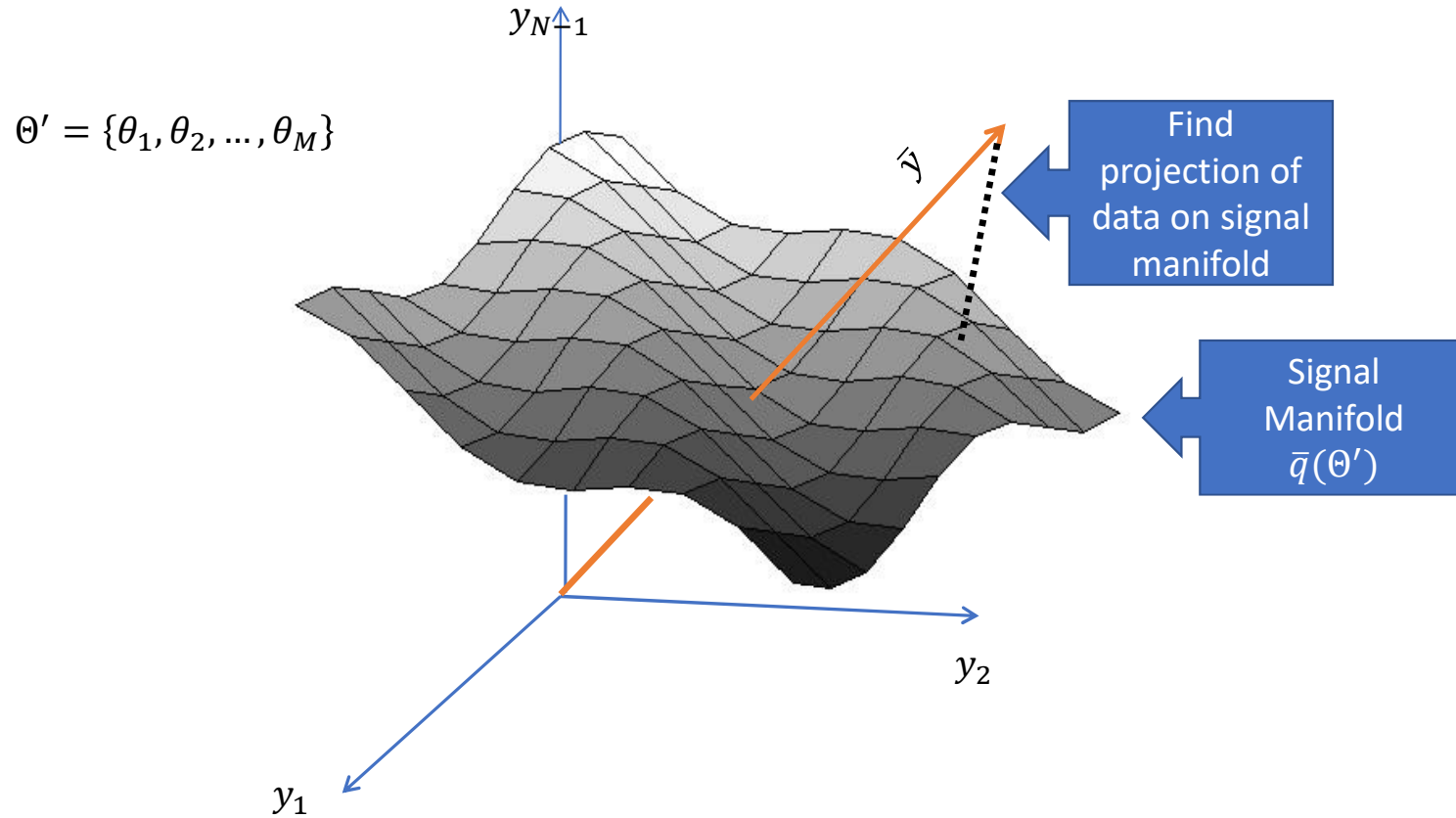
- Large number of maxima
- Becomes worse as the number of parameters increases

Computational cost of matched filtering is generally very high for LIGO data analysis

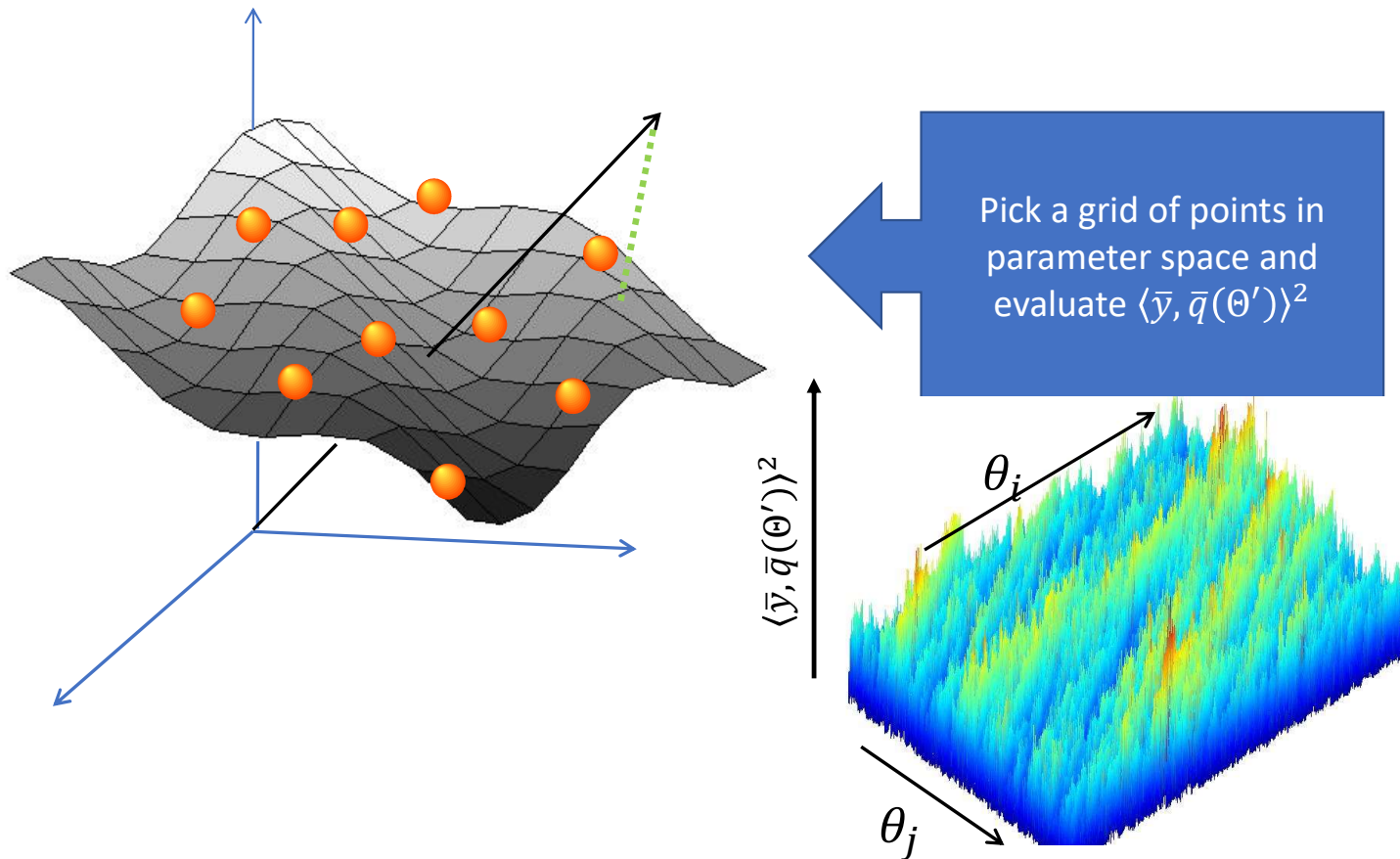
Least squares: general case

- $\|\bar{y} - \bar{s}(\Theta)\|^2 = \|\bar{y}\|^2 - 2\langle \bar{y}, \bar{s}(\Theta) \rangle + \|\bar{s}(\Theta)\|^2$
- $\langle \bar{y}, \bar{s}(\Theta) \rangle = \sum_{i=0}^{N-1} y_i s_i(\Theta)$
 - This is simply the Euclidean inner product of vectors
- One of the parameters that is always present for all GW signals is the signal amplitude (depends on distance r as $\frac{1}{r}$)
- $\bar{s}(A; \Theta') = A\bar{q}(\Theta')$
- To make the definition of A unique, let $\|\bar{q}(\Theta')\| = 1$.
- $\|\bar{y} - \bar{s}(\Theta)\|^2 = \|\bar{y}\|^2 - 2A\langle \bar{y}, \bar{q}(\Theta') \rangle + A^2$
- $\min_{\Theta} \|\bar{y} - \bar{s}(\Theta)\|^2 = \min_{\Theta'} \min_A (\|\bar{y}\|^2 - 2A\langle \bar{y}, \bar{q}(\Theta') \rangle + A^2)$
- Solution of inner minimization: $A = \langle \bar{y}, \bar{q}(\Theta') \rangle$
- $\min_{\Theta} \|\bar{y} - \bar{s}(\Theta)\|^2 = \min_{\Theta'} (-\langle \bar{y}, \bar{q}(\Theta') \rangle)^2 \Rightarrow \max_{\Theta'} \langle \bar{y}, \bar{q}(\Theta') \rangle^2$

Geometrical picture



Grid-based search



Amplitude and initial phase

- $\bar{s}(A, \phi_0; \Theta') = A \bar{q}(\phi_0; \Theta') = AN(\Theta') \sin(\phi(t; \Theta') + \phi_0)$
 - $N(\Theta')$: normalization constant
- $\bar{s}(A, \phi_0; \Theta') = A N(\Theta') \cos(\phi_0) \sin(\phi(t; \Theta')) + AN(\Theta') \sin(\phi_0) \cos(\phi(t; \Theta'))$
- $\bar{s}(A, \phi_0; \Theta') = X \bar{q}_0(\Theta') + Y \bar{q}_1(\Theta')$
- Assume: $\langle \bar{q}_0, \bar{q}_1 \rangle \approx 0$
- $\|\bar{y} - \bar{s}(A, \phi_0; \Theta')\|^2 = \|\bar{y}\|^2 - 2X\langle \bar{y}, \bar{q}_0 \rangle - 2Y\langle \bar{y}, \bar{q}_1 \rangle + X^2 + Y^2$
- $\min_{\Theta'} \min_{A, \phi_0} \|\bar{y} - \bar{s}(A, \phi_0; \Theta')\|^2 = \min_{\Theta'} \min_{X, Y} \|\bar{y} - (X \bar{q}_0(\Theta') + Y \bar{q}_1(\Theta'))\|^2$
- Solution: $X = \langle \bar{y}, \bar{q}_0 \rangle, Y = \langle \bar{y}, \bar{q}_1 \rangle$
- $\max_{\Theta'} [\langle \bar{y}, \bar{q}_0(\Theta') \rangle^2 + \langle \bar{y}, \bar{q}_1(\Theta') \rangle^2]$
- Special case: $\phi(t; \Theta') = \omega t; \quad \Theta' = \omega \rightarrow \max_{\omega} \left| \sum_{k=0}^{N-1} y_k e^{-i\omega k \delta t} \right|^2$
- Discrete Fourier transform: $\tilde{y}_p = \sum_{k=0}^{N-1} y_k e^{-2\pi i p k / N} \approx \left(\frac{1}{\delta t} \right) \int_0^T dt y(t) e^{-2\pi i \left(\frac{p}{T} \right) k \left(\frac{T}{N} \right)} = \left(\frac{1}{\delta t} \right) \int_0^T dt y(t) e^{-i\omega k \delta t} \approx \sum_{k=0}^{N-1} y_k e^{-i\omega k \delta t}$

Amplitude, initial phase, time of arrival

- $\max_{\Theta'} \max_{t_a} [\langle \bar{y}, \bar{q}_0(t - t_a; \Theta') \rangle^2 + \langle \bar{y}, \bar{q}_1(t - t_a; \Theta') \rangle^2]$
 - $\langle \bar{y}, \bar{q}_a(t - t_a; \Theta') \rangle$: Filtering → The whole process is often called “matched filtering” also in GW data analysis
 - Correct name: Least squares fitting
- Example: Linear chirp signal
- $$s(t; A, t_a, f_0, f_1, \phi_0) = \begin{cases} 0; & t \notin [t_a, t_a + L] \\ A N(f_0, f_1) \sin(2\pi(f_0(t - t_a) + f_1(t - t_a)^2) + \phi_0) & \end{cases}$$
- Tasks (Track 1&2):
 - Generate data with this signal and noise
 - Noise: just a sequence of independently drawn values from a Gaussian random number generator
 - Implement matched filtering (whole process) to find this signal

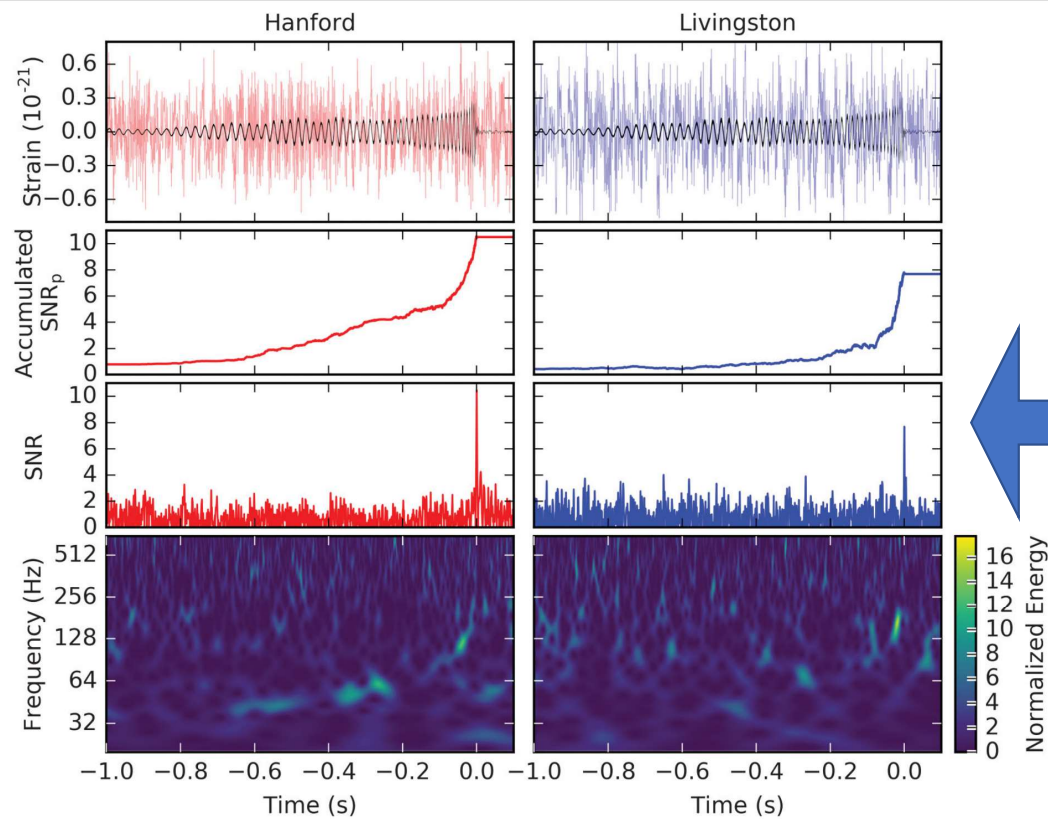
Efficient implementation of filtering

- $\langle \bar{y}, \bar{s}(t_a; \Theta') \rangle = \left(\frac{1}{\delta t} \right) \int_0^T dt y(t) s(t - t_a; \Theta') = C(t_a; \Theta') : \text{Correlation of data with signal}$
- For fixed Θ' , $\min_{t_a} \|\bar{y} - \bar{s}\|^2$ is equivalent to $\max_{t_a} \langle \bar{y}, \bar{s} \rangle$
- Convolution theorem: $F[C(t)](f) = F[y(t)](f) \times F[s(t; \Theta')](f)$
- In terms of Discrete Fourier Transform: $F \rightarrow DFT$
- Inverse DFT: $y_k = \left(\frac{1}{N} \right) \sum_{p=0}^{N-1} \tilde{y}_p e^{2\pi i kp/N}$
- $\sum_{p=0}^{N-1} e^{-2\pi i kp/N} \sum_{l=0}^{N-1} y_l s_{l-p} = \left(\frac{1}{N^2} \right) \sum_{r=0}^{N-1} \sum_{s=0}^{N-1} \tilde{y}_r \tilde{s}_s^* \sum_{p=0}^{N-1} \sum_{l=0}^{N-1} e^{-2\pi i kp/N} e^{2\pi i r l/N} e^{-2\pi i s(l-p)/N}$
- $= \left(\frac{1}{N^2} \right) \sum_{r=0}^{N-1} \sum_{s=0}^{N-1} N^2 \delta_{sr} \delta_{ks} \tilde{y}_r \tilde{s}_s^* = \tilde{y}_k \tilde{s}_k^*$
- (Circular shift of signal: $s_{k<0} = s_{N+k}$)
- Multiply (sample by sample) DFT of data and (complex conjugate) of DFT of signal \rightarrow take inverse DFT
- Use Fast Fourier Transform (FFT) implementation of DFT

GW 151226

PRL 116, 241103 (2016)

PHYSICAL REVIEW LETTERS



Matched filter output
for the filter with best
matching mass
parameters