

Continuous Gravitational Waves: Methods, Results, and Practical Tools

Continuous gravitational waves and neutron stars workshop
AEI Hannover, Germany

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Australian
National
University

Overview

Continuous Wave Sources and Signal Model

Continuous Wave Search Methods

20 Years of Continuous Wave Search Results

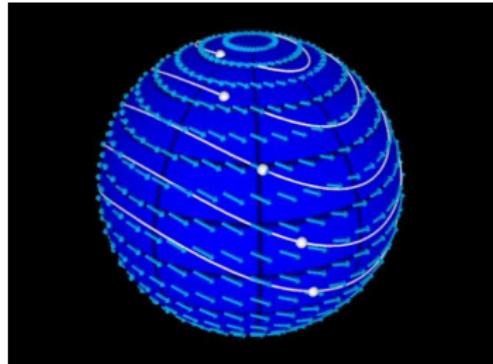
Continuous Wave Software and Practical Examples

Continuous Wave Sources and Signal Model

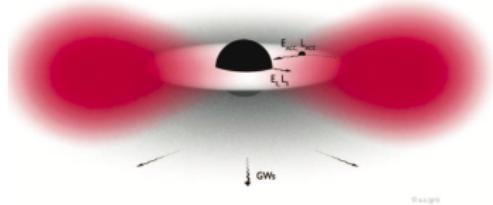
Continuous Wave (CW) Sources



“Mountain” on a neutron star (with a spaceship flying over it and stars in the background, as suggested by Bing AI)



r -modes; current (instead of mass) quadrupole; see Fabian's and Suprovo's talks



Clouds of ultralight boson particles around black holes

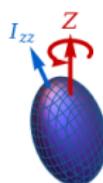
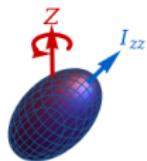
“Mountains” on Neutron Stars (NSs)

Data analysis perspective: NS modelled as classical solid

- ▶ Fine for slowly rotating ($v \ll c$) stars
- ▶ Moment of inertia \mathbf{I} , principal moments I_{xx}, I_{yy}, I_{zz}

3 classes:

1. $I_{xx} = I_{yy} = I_{zz}$: No gravitational waves
2. $I_{xx} = I_{yy} \neq I_{zz}$: Biaxial rotator
 - Ellipticity $\epsilon = |I_{xx} - I_{zz}|/I_{zz}$ assumed small
3. $I_{xx} \neq I_{yy} \neq I_{zz}$: Triaxial rotator
 - Ellipticity $\epsilon = |I_{xx} - I_{yy}|/I_{zz}$ assumed small
 - 2 subclasses:
 - Aligned: a principal axis (z) aligns with rotation axis (Z)
 - Nonaligned: no principal axis aligns with rotation axis



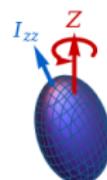
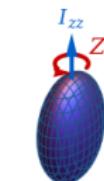
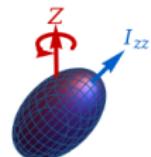
“Mountains” on Neutron Stars (NSs)

CW emission assumed quadrupolar; allowed frequencies:

- ▶ Rotation frequency¹ f_{rot} : NS has no rotational symmetry, rotation about non-principal axis
- ▶ $2f_{\text{rot}}$: NS has 180° rotational symmetry, rotation about principal axis

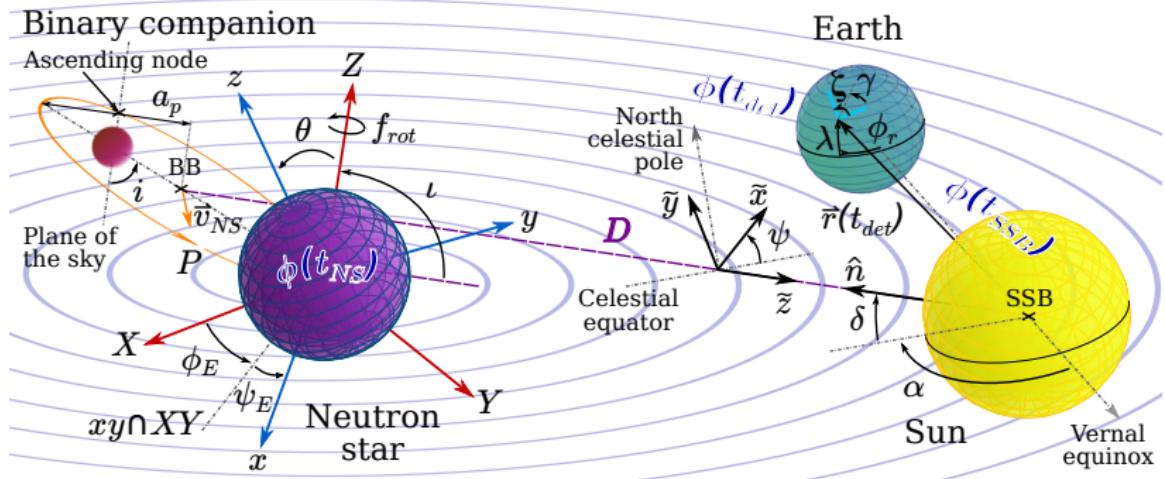
For:

- ▶ $I_{xx} = I_{yy} \neq I_{zz}$: Biaxial rotator
 - CW emission frequencies depends on angle θ between I_{zz} and rotation axis
 - CW emission at rotation frequency f_{rot} maximal at $\theta = 45^\circ$, disappears at $\theta = 90^\circ$
 - CW emission at $2f_{\text{rot}}$ maximal at $\theta = 90^\circ$
- ▶ $I_{xx} \neq I_{yy} \neq I_{zz}$: Triaxial rotator
 - Aligned: CW emission at $2f_{\text{rot}}$ only;
Most commonly assumed by CW searches
 - Nonaligned: CW emission at f_{rot} and $2f_{\text{rot}}$

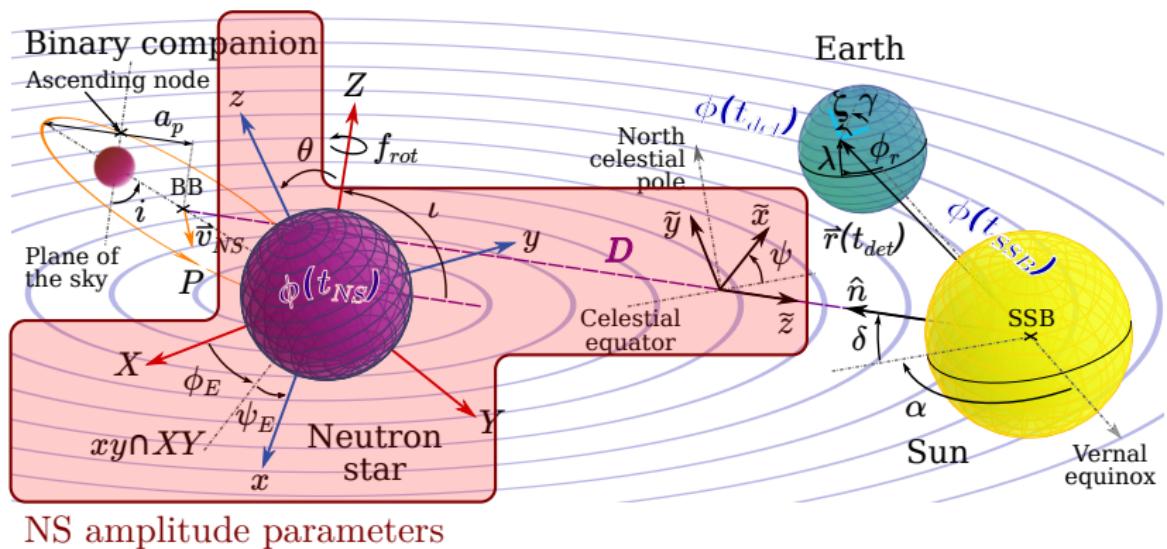


¹possibly plus precession frequency

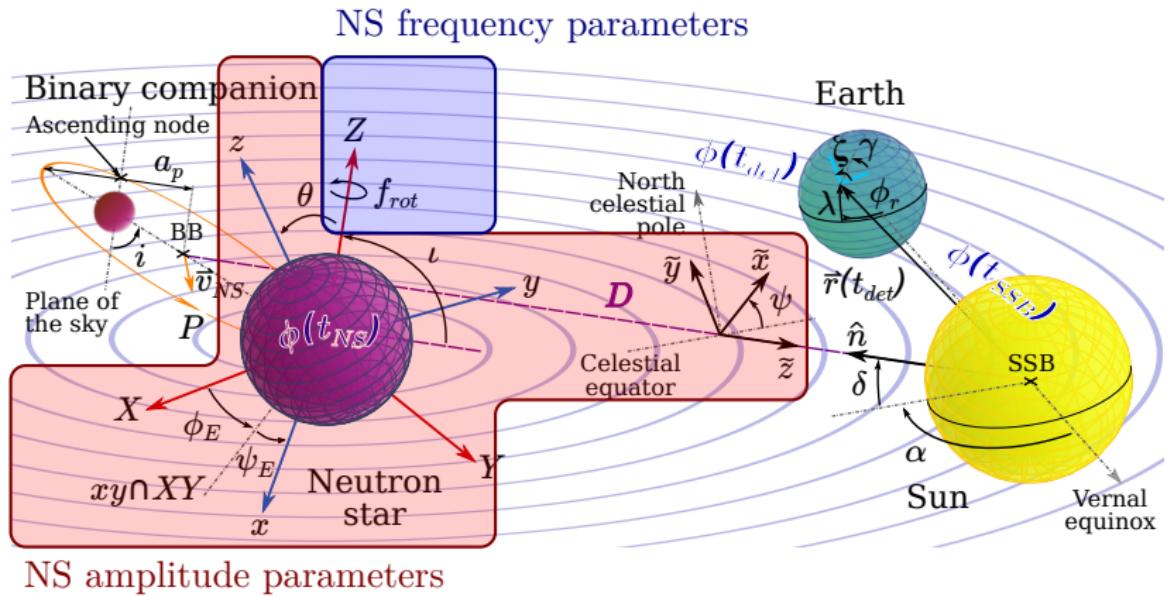
CW Signal Model



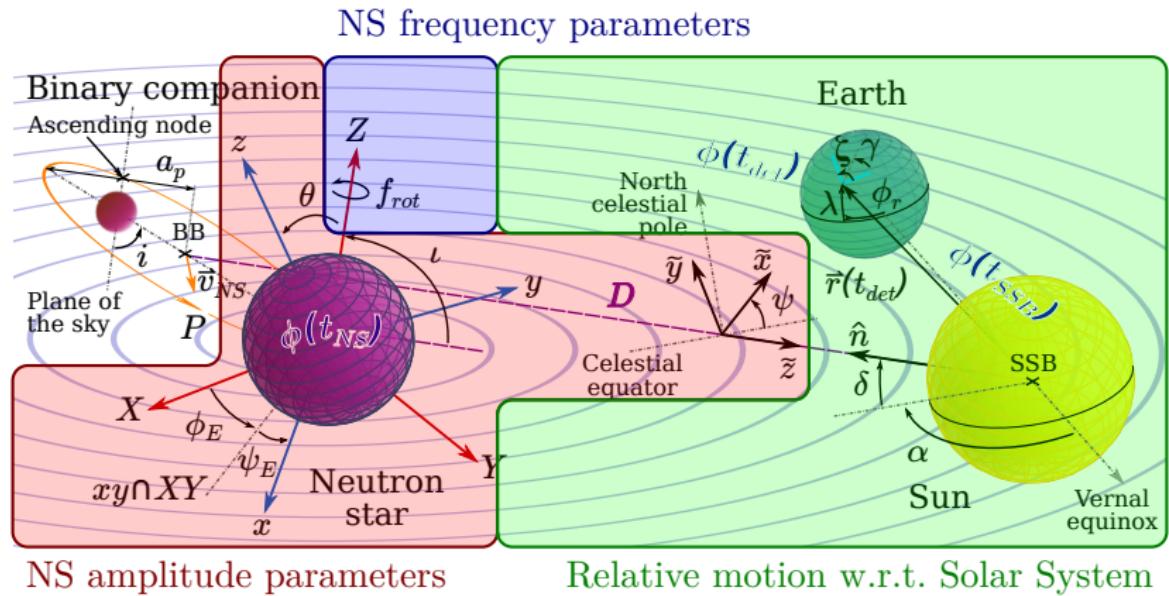
CW Signal Model



CW Signal Model

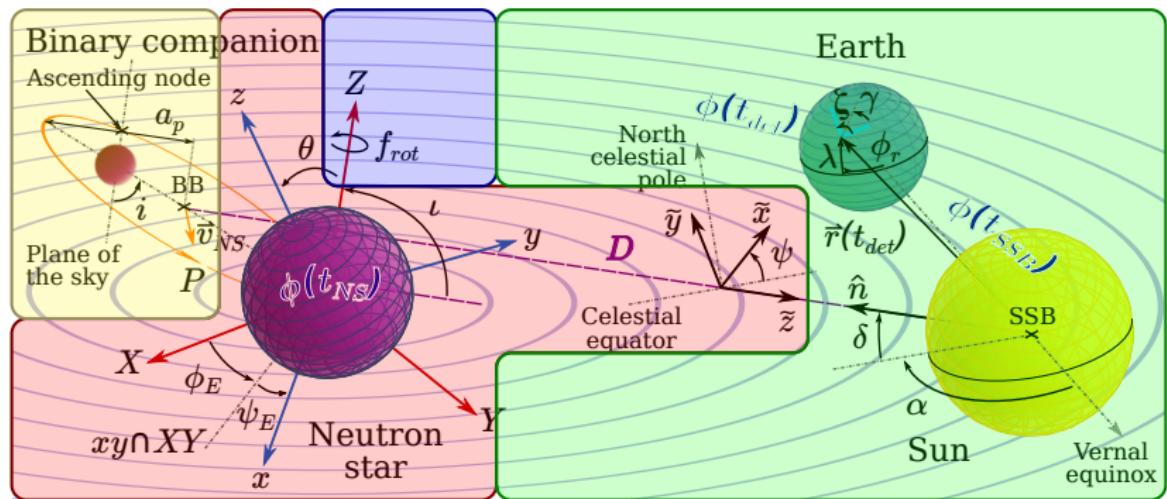


CW Signal Model



CW Signal Model

Relative motion w.r.t. binary companion
NS frequency parameters



CW Signal Model: Amplitude Parameters

- Strength of gravitational wave signal. Triaxial aligned model:

$$h_0 = \frac{4\pi^2 G}{c^4 D} I_{zz} \epsilon f^2, \quad (1)$$

where

- I_{zz} : principal moment
- ϵ : ellipticity
- $f = 2f_{\text{rot}}$: CW frequency
- D : distance

- Orientation of NS relative to detector. Triaxial aligned model:
 - ι : Angle between NS rotation axis and line of sight to detector
 - Controls linear vs. circular polarisation of CW signal
 - ψ : Orientation of basis defining “+”, “ \times ” GW polarisation
 - ϕ_0 : Arbitrary phase of CW signal at reference time
 - ι, ψ may be constrained by EM priors (pulsar wind nebulae)
 - No direct information about NS physics (without EM)
- Data analysis challenge: Easy
 - Find best-fit values from either analytic maximisation (\mathcal{F} -statistic), or Bayesian marginalisation (see Reinhard's talk)

CW Signal Model: Frequency Parameters

- ▶ Evolution of $f(t)$ unknown, but assumed to be slowly varying
- ▶ Can therefore model as Taylor series

$$f(t) = f_0 + f_1(t - t_0) + \frac{1}{2}f_2(t - t_0)^2 + \dots, \quad (2)$$

where

- $f(t)$: Instantaneous CW frequency at time t
 - f_0 : CW frequency at reference time t_0
 - $f_1 \equiv \dot{f}$: 1st *spindown* or frequency time derivative at t_0
 - $f_2 \equiv \ddot{f}$: 2nd *spindown* or frequency time derivative at t_0
 - 2nd spindowns typically only needed for “young” ($\lesssim 1$ kyr) NSs
- ▶ Data analysis challenge: Hard
 - Need to perform “search”: trial-and-error computation of detection statistic for discrete choices of f_0 , f_1 , etc.

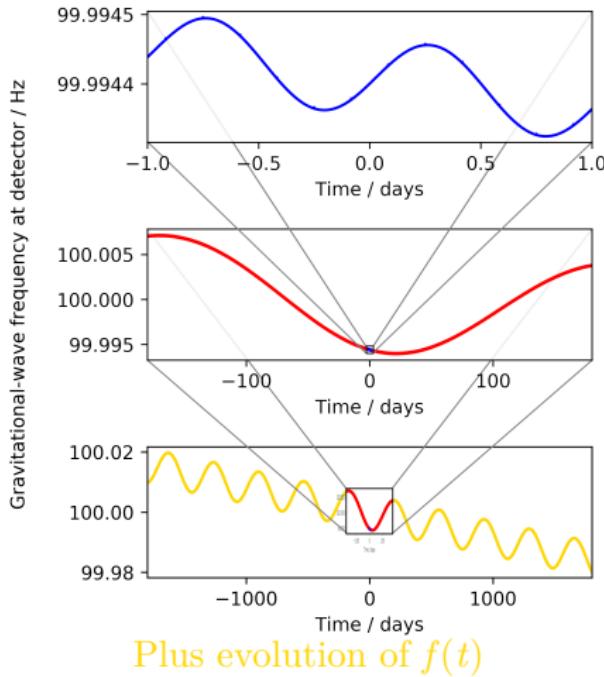
CW Signal Model: Relative Motion

Doppler modulation of CW signal; not a pure sinusoid

- ▶ Solar System: daily rotation of Earth, yearly orbit of Earth around Sun, parameters:
 - right ascension α
 - declination δ
- ▶ Binary companion (not shown), parameters:
 - orbital period P
 - projected semi-major axis a_p
 - if circular orbit: +1 parameter
 - if low-eccentricity orbit: +3 parameters

Data analysis challenge: Hard

- ▶ Need to search over parameters



Continuous Wave Search Methods

GW Signals in Interferometric Detector

Detector measures time series²

$$h(t) = F^{\mu\nu}(t)h_{\mu\nu}(t) \quad (3)$$

$$\begin{aligned} &= A_+(h_0, \cos \iota) F_+(t, \psi) \cos [\phi_0 + \phi(t, \vec{\lambda})] \\ &\quad + A_\times(h_0, \cos \iota) F_\times(t, \psi) \sin [\phi_0 + \phi(t, \vec{\lambda})] \end{aligned} \quad (4)$$

with $A_{+/\times}$ amplitudes, $F_{+/\times}$ antenna responses, ϕ CW signal phase

- ▶ Amplitude parameters (triaxial aligned model): $h_0, \cos \iota, \psi, \phi_0$
- ▶ Phase parameters represented by $\vec{\lambda}$:
 - Evolution of $f(t)$: f_0, f_1, f_2 , etc.
 - Doppler modulation from Earth motion: α, δ
 - Doppler modulation from binary companion: P, a_p , etc.

Reformat Eq. (4) as product of 4 amplitudes \mathcal{A}^μ and 4 functions h_μ ³:

$$h(t) = \mathcal{A}^\mu(h_0, \cos \iota, \psi, \phi_0) h_\mu(t, \vec{\lambda}) \quad (5)$$

²Einstein summation over indices μ, ν in GW tensor $h_{\mu\nu}(t)$ and detector response $F^{\mu\nu}(t)$

³R. Prix, Physical Review D 75, 023004 (2007).

CW Data Analysis: A Solved Problem?

In principle, yes: matched filtering⁴

- ▶ Construct log-likelihood ratio⁵

$$\log \mathcal{L} = \log \frac{P(\text{data } x(t) | x(t) = h(t) + \text{noise})}{P(\text{data } x(t) | x(t) = \text{noise})} \quad (6)$$

$$= (x(t) | h(t)) - \frac{1}{2}(h(t) | h(t)) \quad (7)$$

assuming noise is Gaussian and (quasi-)stationary

- ▶ Functions $h_\mu(t, \vec{\lambda})$ are known for choice of $\vec{\lambda}$
- ▶ Amplitudes \mathcal{A}^μ can be analytically maximised over (\mathcal{F} -statistic)
- ▶ CW search algorithm:
 - Compute $\log \mathcal{L}$ (maximised over \mathcal{A}^μ) for choices of $\vec{\lambda}$
 - Largest value of $\log \mathcal{L}$ found is our most promising candidate
 - Problem solved?!

⁴P. Jaranowski et al., Physical Review D 58, 063001 (1998).

⁵Notation: conditional probability $P(\text{statement} | \text{prior})$; inner product $(x(t) | y(t))$ see Eq. (39) of Jaranowski et al.

CW Data Analysis: A Solved Problem?

In practise: no!!

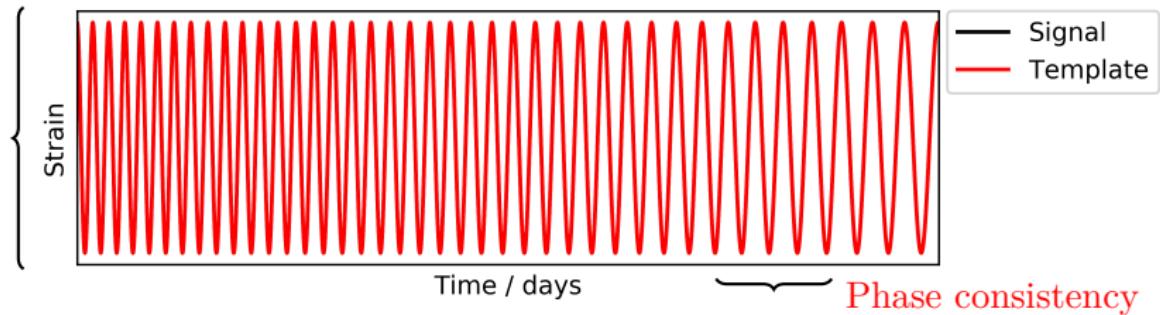
- ▶ Except for CW searches targeting known pulsars, parameter space of possible $\vec{\lambda}$ is veryⁿ large! ($n \gg 1$)
- ▶ Example: search of 1 year of data
 - all-sky survey (i.e. all α, δ)
 - frequency range $f_0 \in [50, 1000]\text{Hz}$
 - 1st spindown $f_1 \in [-10^{-8}, 0]\text{Hz/s}$
 - isolated NS (i.e. no binary orbit parameters)
- ▶ We would need to compute $\log \mathcal{L}$ for approximately
 - ~ 9.9 trillion choices of (α, δ)
 - ~ 40 billion choices of f_0
 - ~ 2.3 million choices of f_1
 - So ~ 920 octillion (10^{27}) values of $\log \mathcal{L}$ in total
- ▶ It takes⁶ ~ 810 microseconds to compute $\log \mathcal{L}$ for 1 year of data and *one* choice of parameters
- ▶ Total search time: ~ 2.3 quintillion (10^{18}) years
 - e.g. ~ 5 billion computers running non-stop for ~ 5 billion years

⁶This assumes the “demodulation” \mathcal{F} -statistic algorithm

The Semi-Coherent Paradigm

Fully coherent filter

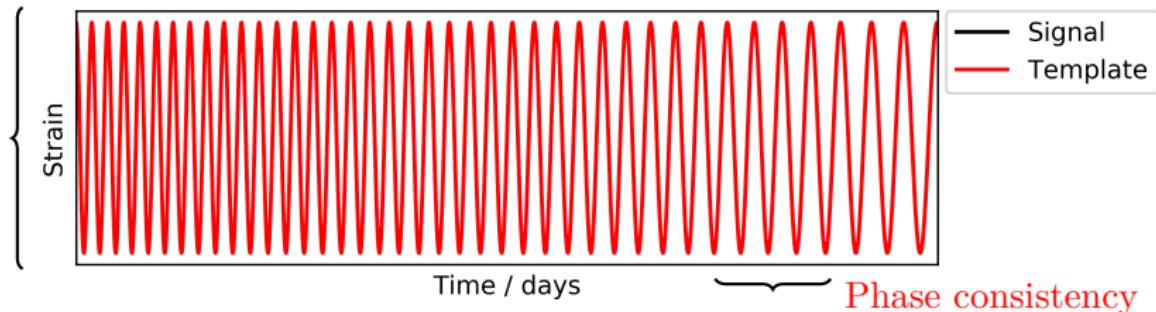
Amplitude consistency



The Semi-Coherent Paradigm

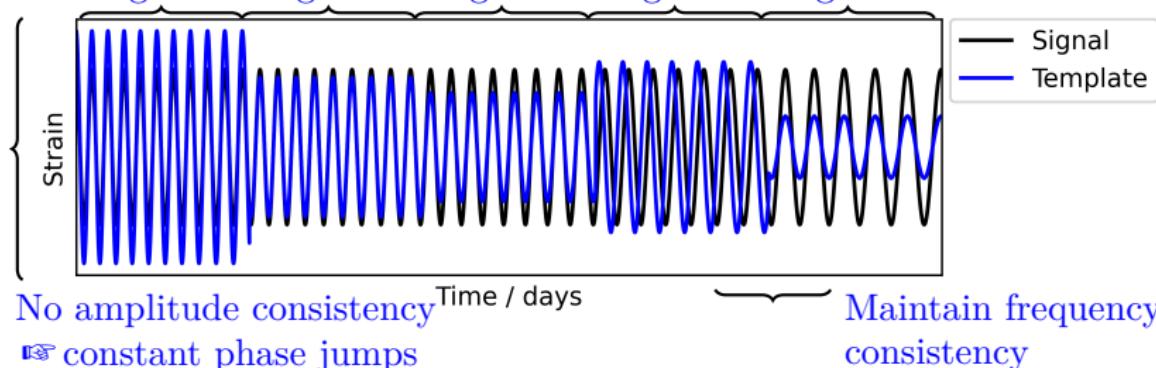
Fully coherent filter

Amplitude consistency



Semi-coherent filter

Segment Segment Segment Segment Segment



The Semi-Coherent Paradigm

► Fully coherent search:

- Template is amplitude and phase consistent with signal
- Computational cost $\propto (\text{observation time})^m$
 - $m \gg 1$
- Sensitivity $\propto (\text{observation time})^{1/2}$

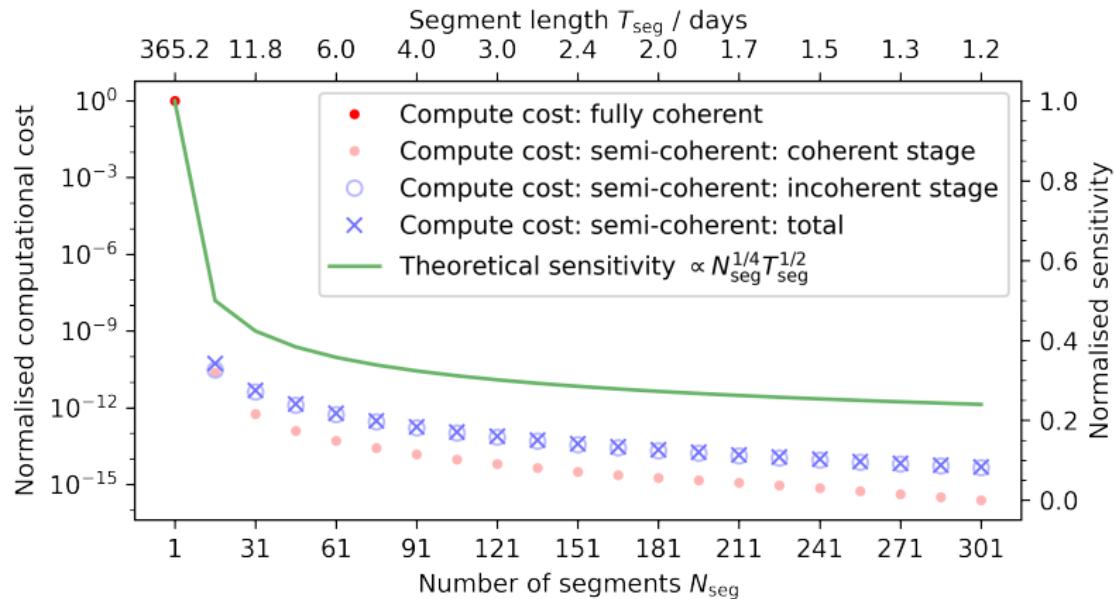
► Semi-coherent search:

- Divide observation time into segments
- Coherent stage:
 - Fully coherent search of each segment individually
 - Amplitude parameters may change between segments
- Incoherent stage:
 - Combine fully-coherent filters from each segment together
 - Maintains frequency consistency over observation
- Computational cost $\propto (\# \text{ of segments})^n (\text{segment time})^m$
 - $(\text{segment time}) \ll (\text{observation time})$
 - $n \ll m^7$
- Sensitivity $\propto (\# \text{ of segments})^{\sim 1/4} (\text{segment time})^{1/2}$

⁷R. Prix and M. Shaltev, Physical Review D 85, 084010 (2012).

The Semi-Coherent Paradigm

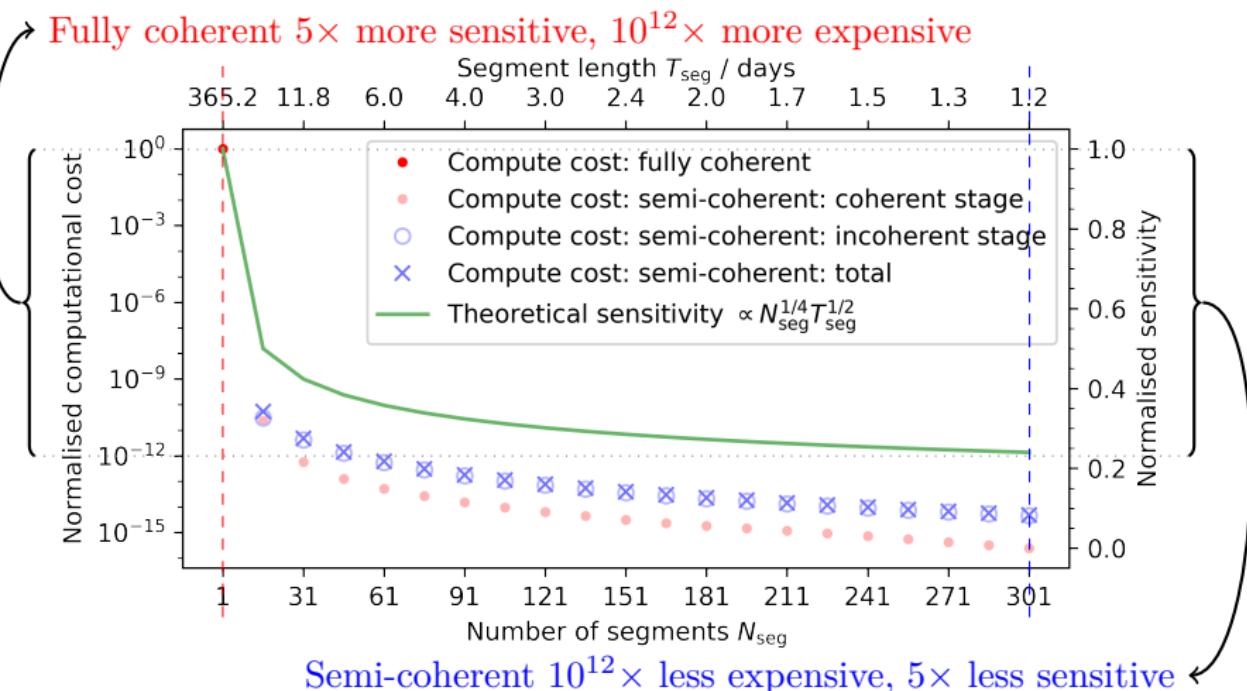
Modelled computational cost of a semi-coherent CW search pipeline⁸ analysing 1 year of data



⁸Weave; K. Wette et al., Physical Review D 97, 123016 (2018)

The Semi-Coherent Paradigm

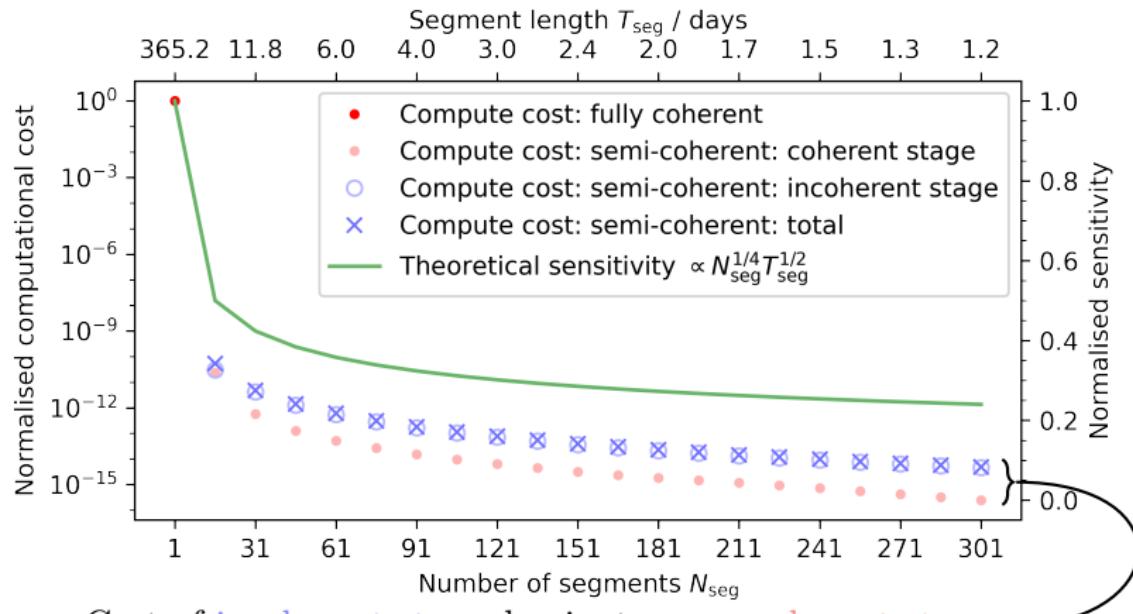
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The Semi-Coherent Paradigm

Modelled computational cost of a semi-coherent CW search pipeline⁸ analysing 1 year of data



⁸Weave; K. Wette et al., Physical Review D 97, 123016 (2018)

Coherent Algorithms

If $T_{\text{seg}} \ll 1 \text{ d}$:

► Frequency domain power:

- Pre-compute Fourier transform of segment data
- Power in a single bin
- Data format/frameworks: Short Fourier Transforms (SFTs)⁹, Short Fourier DataBase (SFDB), Band Sampled Data (BSD)¹⁰

If $T_{\text{seg}} \gtrsim 1 \text{ d}$:

► 5-vector method¹¹:

- Assume all phase parameters are accounted for
- Earth rotation splits CW signal into 5 harmonics
- Sum harmonics at $0, \pm 1, \pm 2$ Earth rotation frequency

► \mathcal{F} -statistic¹²:

- Log-likelihood ratio maximised over \mathcal{A} amplitudes

Fully coherent: heterodyning used for fully-coherent algorithms (e.g. Bayesian inference¹³ over \mathcal{A} amplitudes)

⁹B. Allen et al., tech. rep. T040164-v2 (LIGO, 2022) ¹⁰O. J. Piccinni et al., Classical and Quantum Gravity **36**, 015008 (2019) ¹¹P. Astone et al., Classical and Quantum Gravity **27**, 194016 (2010)

¹²P. Jaranowski et al., Physical Review D **58**, 063001 (1998) ¹³M. Pitkin, Journal of Open Source Software **7**, 4568 (2022)

Incoherent Algorithms

(Loosely) Categorised by detection statistic computed:

- ▶ None:
 - **Coincidence**: self-consistency between segments, inexpensive
- ▶ (Usually) Thresholded statistics:
 - **Hough**¹⁴ family: 2D pattern matching (see Badri's talk)
- ▶ Summed statistics:
 - **StackSlide**¹⁵: unweighted, often used with \mathcal{F} -statistic
 - Algorithms for combining self-consistent \mathcal{F} -statistics:
 - **GCT**¹⁶: based on studying large-scale correlations
 - **Weave**¹⁷: parameter space metric, optimal lattices
 - **PowerFlux**¹⁸: weighted by antenna response functions

¹⁴B. Krishnan et al., Physical Review D **70**, 082001 (2004) ¹⁵P. R. Brady and T. Creighton, Physical Review D **61**, 082001 (2000) ¹⁶H. J. Pletsch and B. Allen, Physical Review Letters **103**, 181102 (2009) ¹⁷K. Wette et al., Physical Review D **97**, 123016 (2018) ¹⁸V. Dergachev, tech. rep. T1000272-v5 (LIGO, 2011)

Incoherent Algorithms

(Loosely) Categorised by signal model assumed:

- ▶ Tunable coherence: correlate segments at different times
 - Loosely coherent¹⁹: all-sky surveys
 - Cross-correlation²⁰: directed targets esp. Scorpius X-1
- ▶ Stochastic frequency wandering:
 - Viterbi²¹: solves Hidden Markov Model
 - SOAP²²: tunable memory
- ▶ Binary specialists:
 - TwoSpect²³: looks for patterns in spectrograms
 - BinarySkyHough²⁴: extension of Hough method
 - BinarySkyHou \mathcal{F} ²⁵: uses \mathcal{F} -statistic
 - BinaryWeave²⁶: extension of Weave; see Arunava's talk

¹⁹V. Dergachev, Classical and Quantum Gravity **27**, 205017 (2010) ²⁰J. T. Whelan et al., Physical Review D **91**, 102005 (2015) ²¹S. Suvorova et al., Physical Review D **93**, 123009 (2016) ²²J. Bayley et al., Physical Review D **100**, 023006 (2019) ²³E. Goetz and K. Riles, Classical and Quantum Gravity **28**, 215006 (2011) ²⁴P. B. Covas and A. M. Sintes, Physical Review D **99**, 124019 (2019)
²⁵P. B. Covas and R. Prix, Physical Review D **106**, 084035 (2022) ²⁶A. Mukherjee et al., Physical Review D **107**, 062005 (2023)

\mathcal{F} -statistic

- Log-likelihood ratio²⁷

$$\log \mathcal{L} = (x(t) | h(t)) - \frac{1}{2} (h(t) | h(t)) \quad (8)$$

$$= \vec{\mathcal{A}} \cdot \vec{x} - \frac{1}{2} \vec{\mathcal{A}} \cdot \mathcal{M} \cdot \vec{\mathcal{A}} \quad (9)$$

where $x_\mu = (x(t) | h_\mu(t))$, $\mathcal{M}_{\mu\nu} = (h_\mu(t) | h_\nu(t))$

- $\log \mathcal{L}$ is quadratic in $\vec{\mathcal{A}}$, analytic maximisation:

$$\frac{\partial \log \mathcal{L}}{\partial \mathcal{A}} = 0 \text{ is solved by } \vec{\mathcal{A}} = \vec{\mathcal{A}}_{\text{MLE}} = \mathcal{M}^{-1} \cdot \vec{x} \quad (10)$$

$$2\mathcal{F} = 2 \log \mathcal{L}|_{\vec{\mathcal{A}}=\vec{\mathcal{A}}_{\text{MLE}}} = \vec{x} \cdot \mathcal{M}^{-1} \cdot \vec{x} \quad (11)$$

- Note: despite being named \mathcal{F} -statistic²⁸, value of statistic is always denoted **2F**

²⁷Notation: (\cdot) represents contraction over neighbouring vector/matrix dimensions

²⁸P. Jaranowski et al., Physical Review D 58, 063001 (1998).

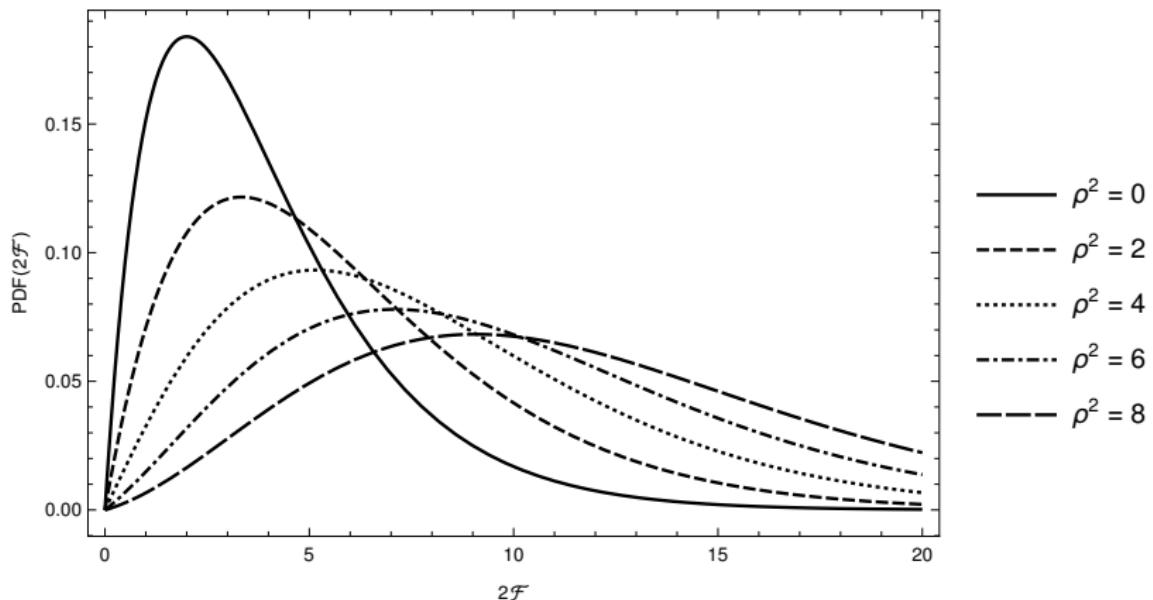
\mathcal{F} -statistic

Probability distribution:

- ▶ Can show \vec{x} follows a multivariate normal distribution with means $\vec{\mu} = \mathbf{M} \cdot \vec{\mathcal{A}}$ and covariance matrix \mathbf{M}
 - $h_\mu(t)$ are not mutually orthogonal
- ▶ Cholesky decomposition $\mathbf{W} \cdot \mathbf{W}^T = \mathbf{M}^{-1}$
- ▶ $\vec{y} = \mathbf{W}^T \cdot \vec{x}$ are uncorrelated normal variables with means $\vec{\nu} = \mathbf{W}^T \cdot \vec{\mu}$ and unit variances
- ▶ $2\mathcal{F} = \vec{x} \cdot \mathbf{M}^{-1} \cdot \vec{x} = \|\vec{y}\|^2$ follows a χ^2 distribution with 4 degrees of freedom
 - y_μ are normal, y_μ^2 are χ^2 with 1 degree of freedom
 - Noncentrality parameter $\rho^2 = \|\vec{\nu}\|^2 = \vec{\mathcal{A}} \cdot \mathbf{M} \cdot \vec{\mathcal{A}}$ is optimal signal-to-noise ratio of CW signal

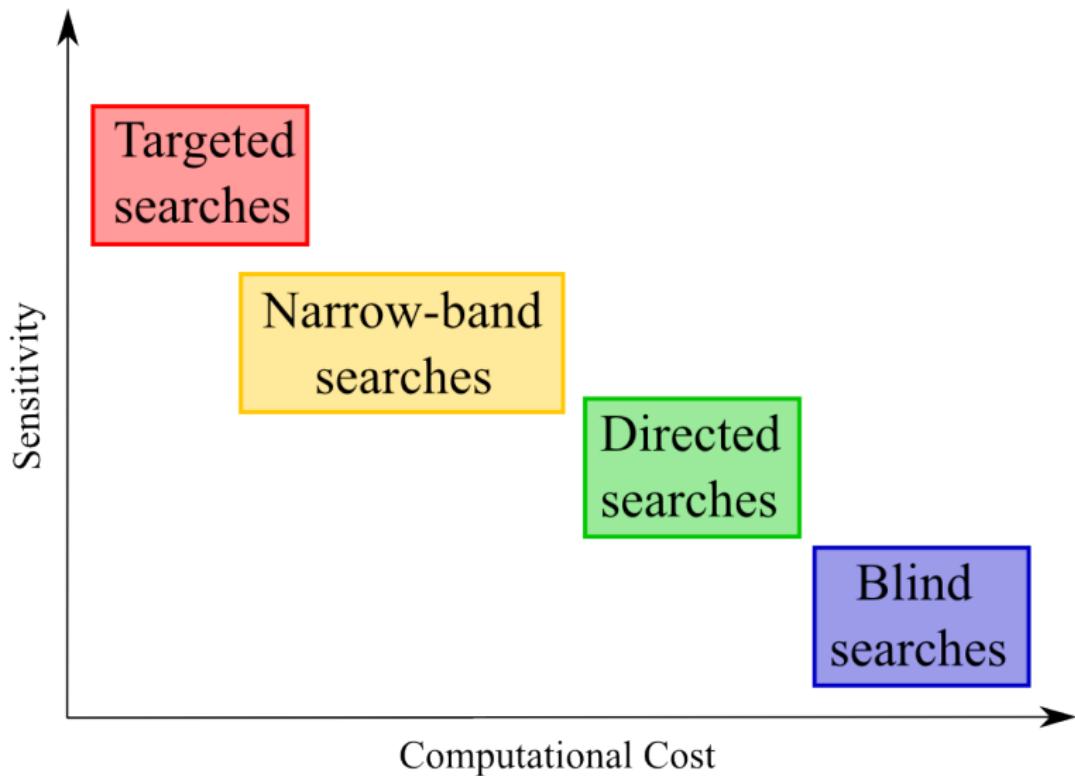
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Probability distribution:

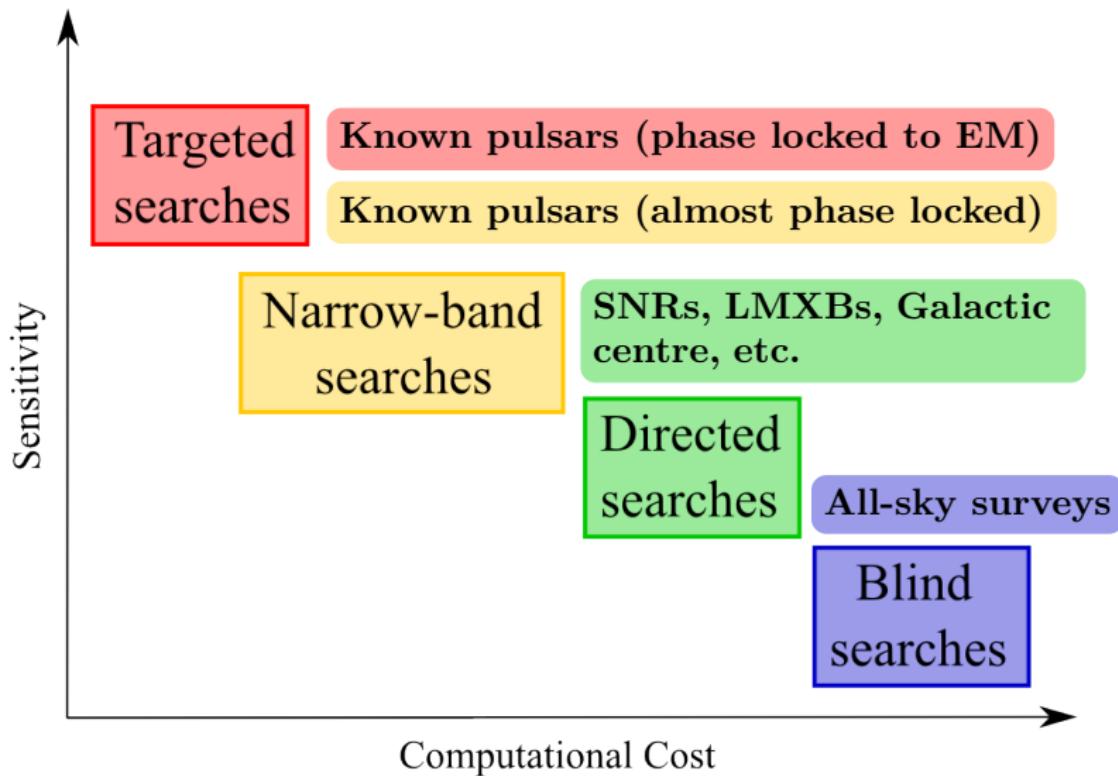


20 Years of Continuous Wave Search Results

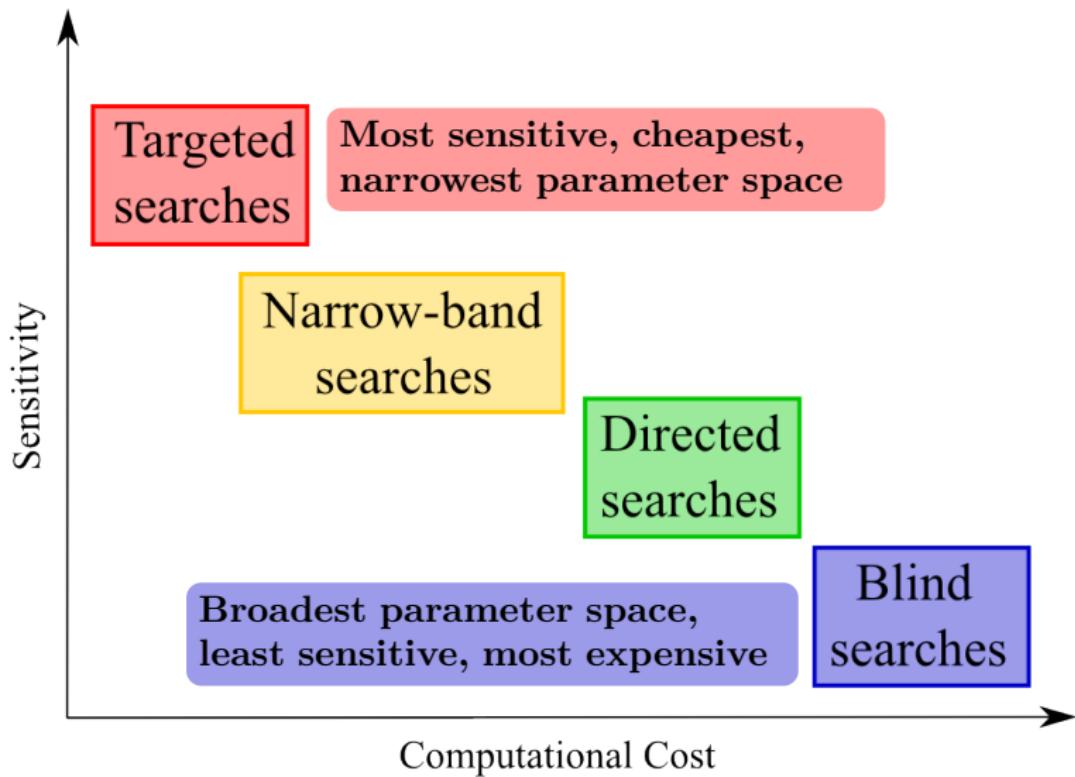
Sensitivity vs Parameter Space/Computational Cost



Sensitivity vs Parameter Space/Computational Cost



Sensitivity vs Parameter Space/Computational Cost



Sensitivity vs Parameter Space/Computational Cost

297 CW searches from 80 published articles (2003–2022)

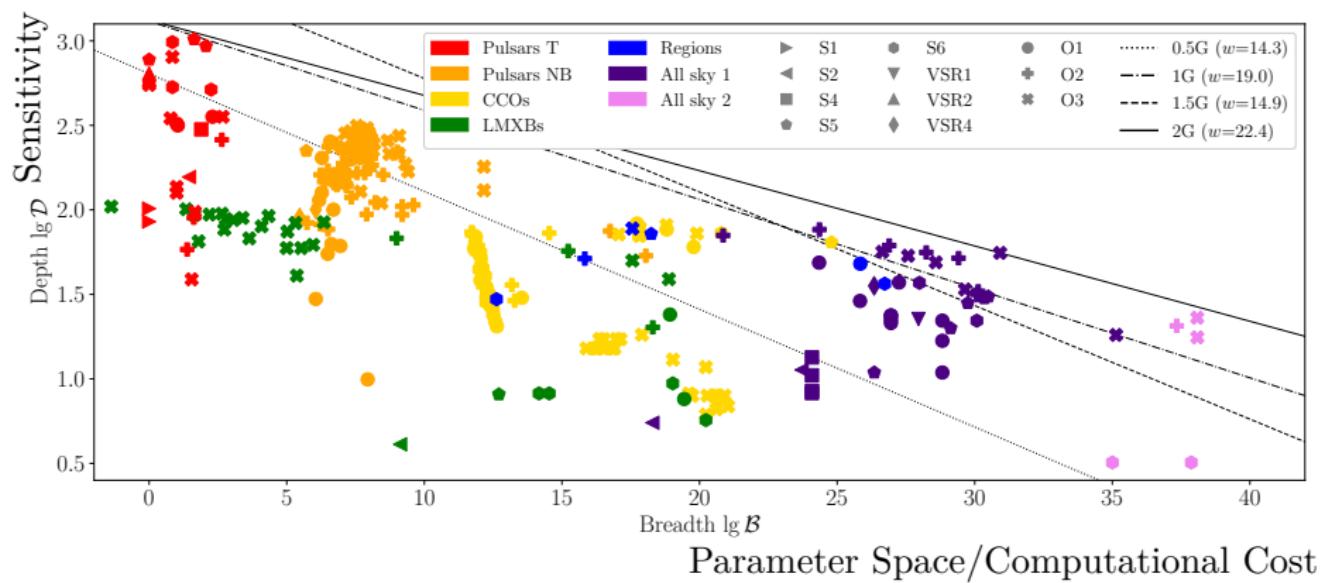
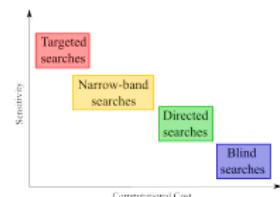
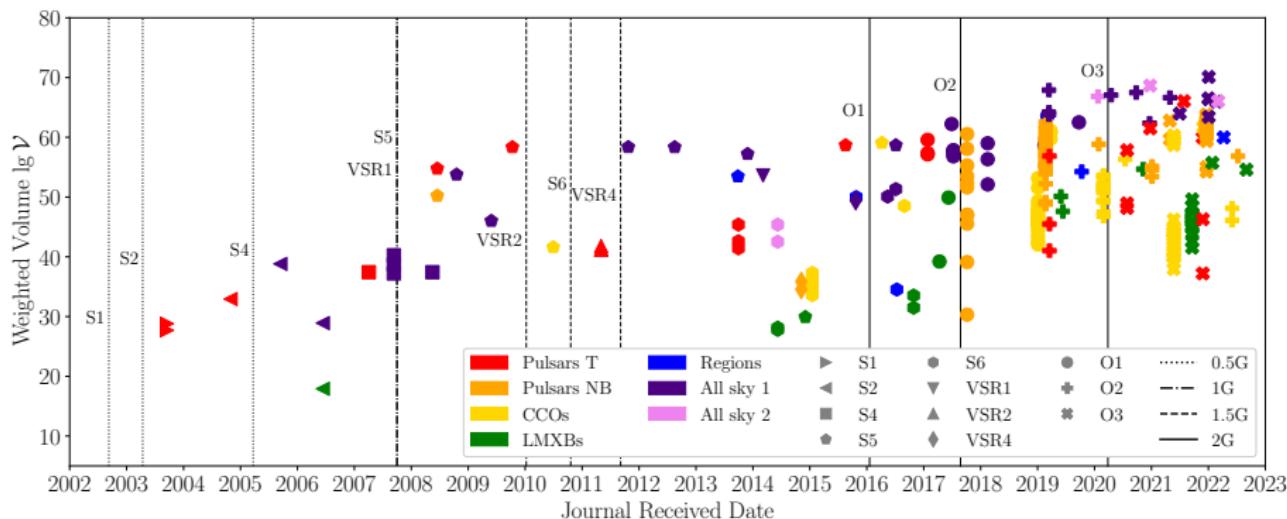


Image credits: K. Wette, Astroparticle Physics 153, 102880 (2023), M. Sieniawska and M. Bejger, Universe 5, 217 (2019)

Sensitivity–Parameter Space Volume

Weighted according to trend during 4 eras:

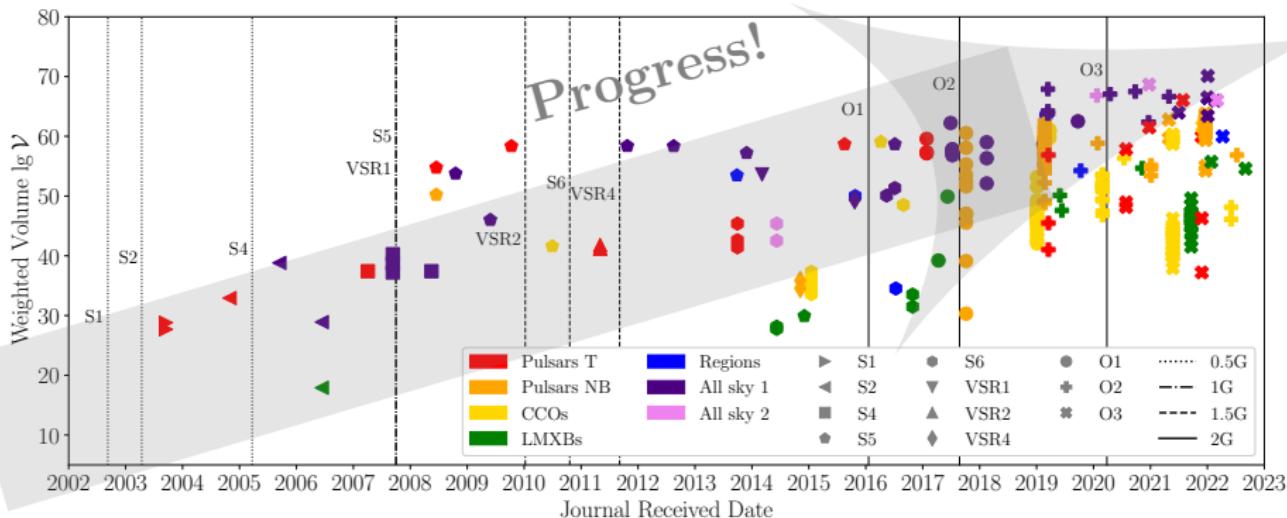
- ▶ 0.5G: pre-initial LIGO (S1–4)
- ▶ 1G: initial LIGO (S5), Virgo (VSR1)
- ▶ 1.5G: enhanced LIGO (S5), Virgo (VSR2–4)
- ▶ 2G: advanced LIGO/Virgo (O1+); see David's talk for LVK O4 plans



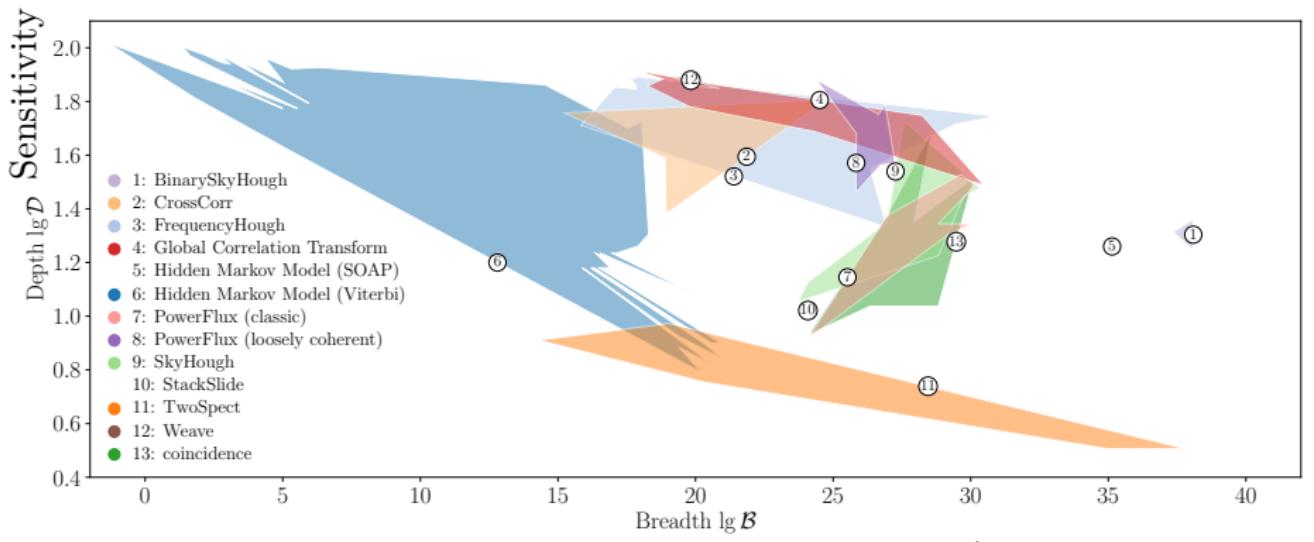
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Distribution of Algorithms



Continuous Wave Software and Practical Examples

LALSuite and CW Signal Examples

LIGO-Virgo-KAGRA Algorithm Library Suite (LALSuite)²⁹:

- ▶ Large collection of libraries for GW data analysis, including:
 - LAL: core routines
 - LALPulsar: routines and applications for CW data analysis
- ▶ Pros:
 - Mature, packaged product: install using Pip/Conda
 - Accessible from Python through SWIGLAL interface³⁰
 - API-level documentation³¹ is generally pretty good
- ▶ Cons:
 - Written in C, hard to hack to do new things
 - Tutorial-level documentation largely absent

CW Signal Examples:

- ▶ Generate a traditional CW signal
- ▶ Generate a CW signal with a user-defined frequency evolution
- ▶ **Link to IPython notebook in Google Colab**

²⁹LIGO Scientific Collaboration et al., Free software (GPL), 2018, <https://doi.org/10.7935/GT1W-FZ16>

³⁰K. Wette, SoftwareX 12, 100634 (2020) ³¹<https://lscsoft.docs.ligo.org/lalsuite/>