

Bayesian \mathcal{F} -statistic-based parameter estimation of continuous gravitational waves from known pulsars

Based on A. Ashok et al., PRD 109, 104002 (2024)

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Outline

- 1 Introduction
- 2 Likelihoods and Bayesian framework
- 3 Tests
- 4 Search for PSR J1526-2744
- 5 Conclusions

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Targeted CW searches

Types of CWs searches using data from ground-based detectors like Advanced LIGO:



- Massive reduction in parameter-space size compared to wide parameter-space searches: a targeted search can use a fully coherent combination of all the data, **leading to the maximum possible sensitivity**
- Presence of a neutron star is assured and its rotational frequency and spin-down are known: **a null measurement is directly informative about the gravitational-wave emission of the source**

Motivation for a new pipeline

- We propose a new pipeline to carry out targeted searches
- Previously, **only a single Bayesian method existed** for parameter estimation on known pulsars, often referred to as the Time Domain method or Heterodyne method
- We introduce a new expression and implementation of the CW signal likelihood function, based on the **well-established and efficient** \mathcal{F} -statistic framework
- Instead of analytically maximizing the likelihood ratio over the 4 amplitude parameters, we set up priors on them and keep the complete likelihood or analytically marginalize over ϕ_0
- This approach can be used for any combination of parameters describing the CW signal, but here we focus on targeted searches, **only computing the posterior over the unknown amplitude parameters**

CW signal model

- Varying mass quadrupole moment in a fast-rotating neutron star. In the absence of precession, the **signal is expected at twice the rotation frequency** and with twice the rotational spin-down of the star.
- The signal strain s in the detector X has the form

$$s^X(t) = F_+(t; \alpha, \delta, \psi) h_+(t) + F_\times(t; \alpha, \delta, \psi) h_\times(t),$$

where “+” and “ \times ” indicate the two gravitational-wave polarizations, and $F_+(t; \alpha, \delta, \psi)$ and $F_\times(t; \alpha, \delta, \psi)$ are the detector antenna-pattern functions. The two waveforms $h_+(t)$ and $h_\times(t)$ are given by

$$h_+(t) = \frac{h_0}{2} (1 + \cos^2 \iota) \cos [\phi(t) + \phi_0],$$

$$h_\times(t) = h_0 \cos \iota \sin [\phi(t) + \phi_0].$$

Amplitude and phase-evolution parameters

Two different types of **parameters** define the CW signal:

Amplitude A

- Gravitational-wave amplitude:

$$h_0 = C \frac{I_{zz} \epsilon f_0^2}{d}$$

- Inclination angle: ι
- Polarization angle: ψ
- Initial phase: ϕ_0

Phase-evolution λ

- Frequency evolution: f_0, f_1, \dots
- Sky position: α and δ
- Keplerian binary orbit: a_p (projected semi-major axis), P_{orb} (orbital period), t_{asc} (time of ascension), e (eccentricity), and ω (argument of periapsis)

In a targeted search, **all phase-evolution parameters are known** or uncertainty is small enough to allow for a fully coherent search

Signal model in JKS amplitude parameters

CW signal depends non-linearly on the physical amplitude parameters \mathcal{A} . A different set of amplitude parameters \mathcal{A} linearize the CW signal $s^X(t)$ in the frame of detector X :

$$s^X(t; \mathcal{A}, \lambda) = \sum_{\mu=1}^4 \mathcal{A}^\mu h_\mu^X(t; \lambda),$$

where h_μ^X are basis functions (depend on phase-evolution parameters) and

$$\mathcal{A}^1 \equiv \frac{h_0(1 + \cos^2 \iota)}{2} \cos \phi \cos \psi - h_0 \cos \iota \sin \phi \sin \psi,$$

$$\mathcal{A}^2 \equiv \frac{h_0(1 + \cos^2 \iota)}{2} \cos \phi \sin \psi + h_0 \cos \iota \sin \phi \cos \psi,$$

$$\mathcal{A}^3 \equiv -\frac{h_0(1 + \cos^2 \iota)}{2} \sin \phi \cos \psi - h_0 \cos \iota \cos \phi \sin \psi,$$

$$\mathcal{A}^4 \equiv -\frac{h_0(1 + \cos^2 \iota)}{2} \sin \phi \sin \psi + h_0 \cos \iota \cos \phi \cos \psi$$

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Likelihood ratio I

We define the scalar product:

$$(x|y) \equiv 2 \sum_X^{N_{\text{det}}} S_X^{-1} \int_0^T x^X(t) y^X(t) dt$$

- The noise hypothesis \mathcal{H}_N states that the strain data $x^X(t)$ contains only (Gaussian) noise $n^X(t)$, i.e., $s^X = 0$. It can be shown that the likelihood function for \mathcal{H}_N is:

$$P(x|\mathcal{H}_N) = \kappa e^{-\frac{1}{2}(x|x)}$$

- The signal hypothesis \mathcal{H}_S states that the strain data $x^X(t)$ contains a signal $s^X(t)$ in addition to (Gaussian) noise $n^X(t)$:

$$x^X(t) = n^X(t) + s^X(t; \mathcal{A}, \lambda)$$

$$P(x|\mathcal{H}_S, \mathcal{A}, \lambda) = \kappa e^{-\frac{1}{2}(x-s|x-s)}$$

Likelihood ratio II

To decide whether the signal or noise hypothesis is favored by the data x , both frequentist and Bayesian frameworks **require expressing the likelihood ratio between the two hypotheses \mathcal{L}** :

$$\mathcal{L}(x; \mathcal{A}, \lambda) \equiv \frac{P(x|\mathcal{H}_S, \mathcal{A}, \lambda)}{P(x|\mathcal{H}_N)} = e^{(x|s) - \frac{1}{2}(s|s)},$$

which when substituting the signal $s^X = \sum_{\mu=1}^4 \mathcal{A}^\mu h_\mu^X(t; \lambda)$ becomes:

$$\log \mathcal{L}(x; \mathcal{A}, \lambda) = \mathcal{A}^\mu x_\mu - \frac{1}{2} \mathcal{A}^\mu \mathcal{M}_{\mu\nu} \mathcal{A}^\nu = \mathcal{A}^\mu x_\mu - \frac{\rho^2}{2},$$

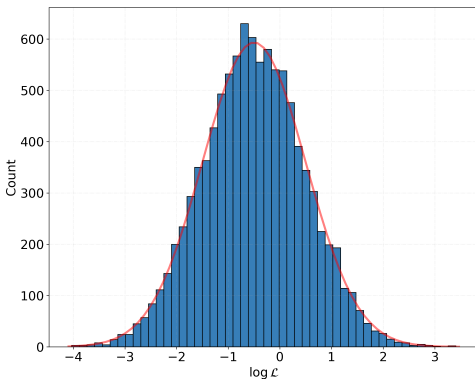
with implicit summation over repeated amplitude indices $\mu, \nu = 1 \dots 4$, and the definitions

$$x_\mu(\lambda) \equiv (x|h_\mu) \quad \text{and} \quad \mathcal{M}_{\mu\nu}(\lambda) \equiv (h_\mu|h_\nu)$$

This likelihood is **mathematically equivalent to the one used by the Time-domain heterodyning method**, although its calculation is done differently

Statistical properties

$\log \mathcal{L}$ is a **Gaussian-distributed quantity** since each x_μ follows a Gaussian distribution.



Histogram of 10 000 log-likelihood ratio values for the noise-only case. The red line shows the expected Gaussian distribution with a mean of -0.5 and a standard deviation of 1.

Marginalized likelihood I

The likelihood ratio \mathcal{L} **can be analytically marginalized over** ϕ_0 . Some advantages:

- It leaves us with less dimensions to explore numerically (4 to 3)
- It avoids the bi-modality of posteriors in $\psi - \phi_0$
- It provides a consistency check for the results from the full-likelihood

We can **explicitly factor out the ϕ_0 -dependence** in the $\mathcal{A}^\mu x_\mu$ term:

$$\mathcal{A}^\mu x_\mu = q \cos(\phi_0 - \varphi_0),$$

where q and φ_0 are independent of ϕ_0 .

Marginalized likelihood II

We can see that the signal power ρ^2 does not depend on ϕ_0 , and therefore writing the likelihood ratio in the form

$$\mathcal{L}(x; \mathcal{A}) = e^{-\frac{1}{2}\rho^2} e^{q \cos(\phi_0 - \varphi_0)}$$

makes the ϕ_0 dependence fully explicit. Using a uniform ϕ_0 -prior we can **obtain the ϕ_0 -marginalized likelihood ratio $\mathcal{L}^{\overline{\phi_0}}$** in the form

$$\begin{aligned} \mathcal{L}^{\overline{\phi_0}} &\equiv \int_0^{2\pi} \mathcal{L}(x; \mathcal{A}) P(\phi_0 | \mathcal{H}_S) d\phi_0 \\ &= \frac{1}{2\pi} \int_0^{2\pi} \mathcal{L}(x; \mathcal{A}) d\phi_0 \\ &= \frac{1}{2\pi} e^{-\frac{1}{2}\rho^2} \int_0^{2\pi} e^{q \cos(\phi_0 - \varphi_0)} d\phi_0 = e^{-\frac{1}{2}\rho^2} I_0(q), \end{aligned}$$

where we used the Jacobi-Anger expansion to see that $\int_0^{2\pi} e^{q \cos \phi} d\phi = 2\pi I_0(q)$ in terms of the modified Bessel function of the first kind I_0 .

Priors on amplitude parameters I

If there are **no observational constraints on the rotation axis of the pulsar**, we assume isotropic “ignorance” priors on the angle parameters:

- The initial phase ϕ_0 corresponds to the pulsar rotation angle at a reference time and the ignorance prior is uniform over the range $\phi_0 \in [0, 2\pi)$
- The ignorance prior for the direction of the rotation axis is also uniform $\in [0, 2\pi)$ and it translates to uniform priors in $\cos \iota \in [-1, 1]$ and $\psi \in [0, 2\pi)$
- We see that $\psi \rightarrow \psi + \pi$ leaves the \mathcal{A}^μ unchanged, and further that $\psi \rightarrow \psi + \pi/2$ flips their sign, which can be compensated by $\phi_0 \rightarrow \phi_0 + \pi$. We can therefore choose a gauge where $\psi \in [-\pi/4, \pi/4)$ and $\phi_0 \in [0, 2\pi)$.

When pulsar observations do constrain these priors, they can be modified appropriately.

Priors on amplitude parameters II

For h_0 the choice of prior range $[h_{\text{low}}, h_{\text{high}}]$ and probability distribution is less straightforward:

- For a known pulsar, one could inform h_{high} from the observed pulsar parameters, namely the spindown upper limit $h_{\text{high}} < h_0^{\text{sd}}$
- If a previous search has established h_0^{UL} for the pulsar, then, under the assumption that the signal amplitude does not change over time, one could require $h_{\text{high}} < h_0^{\text{UL}}$
- Use theoretical range of ellipticities ε of neutron stars to derive an h_0 prior range from $h_0(\varepsilon) = \frac{4\pi^2 G \varepsilon I_{zz} f^2}{c^4 d}$
- Quadrupolar deformations can be sourced by magnetic field B : smallest expected signal may correspond to this, so $h_{\text{low}} \sim h_0(B)$

For a strong signal, the prior has minimal influence on the resulting posterior (likelihood strongly peaked). For a weak signal, a uniform prior on h_0 leads to a more conservative upper limit compared to a log-uniform distribution. Conversely, a log-uniform prior ensures a uniform sampling when our ignorance spans several orders of magnitude.

Bayes' theorem

Using Bayes' theorem, **the posterior for the unknown amplitude parameters \mathcal{A} is:**

$$\begin{aligned} P(\mathcal{A}|x, \mathcal{H}_S, \lambda) &= P(\mathcal{A}|\mathcal{H}_S, \lambda) \frac{P(x|\mathcal{H}_S, \mathcal{A}, \lambda)}{P(x|\mathcal{H}_S, \lambda)} \\ &= P(\mathcal{A}|\mathcal{H}_S, \lambda) \frac{\mathcal{L}(x; \mathcal{A}, \lambda) P(x|\mathcal{H}_N)}{P(x|\mathcal{H}_S, \lambda)} \end{aligned}$$

where $P(\mathcal{A}|\mathcal{H}_S, \lambda)$ is the prior on the amplitude parameters, $P(x|\mathcal{H}_S, \mathcal{A}, \lambda)$ is the signal likelihood, and $P(x|\mathcal{H}_S, \lambda)$ is the amplitude-marginalized signal likelihood.

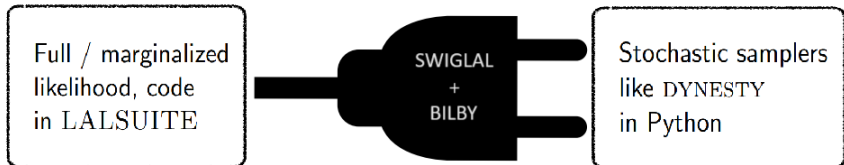
Collecting all \mathcal{A} -independent factors into a proportionality constant k :

$$P(\mathcal{A}|x, \mathcal{H}_S, \lambda) = k \mathcal{L}(x; \mathcal{A}, \lambda) P(\mathcal{A}|\mathcal{H}_S, \lambda)$$

where k can be determined via the normalization condition

$$\int P(\mathcal{A}|\dots) d^4\mathcal{A} = 1.$$

Software



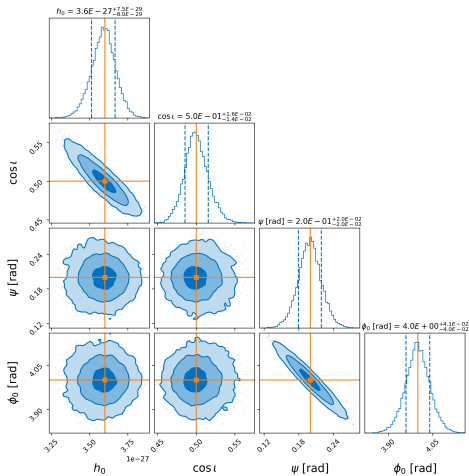
- Two stochastic samplers tested: DYNESTY and BILBYMCMC
- On an Intel(R) Xeon(R) CPU E5-2620 v4 @ 2.10 GHz processor, the median time for a single computation of the likelihood function is $\sim \mathcal{O}(\mu s)$
- The ϕ_0 -marginalized likelihood needs around twice as the full likelihood
- The known-pulsar search we did took 154 seconds (with 32-core parallelization) with the full likelihood

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Tests I

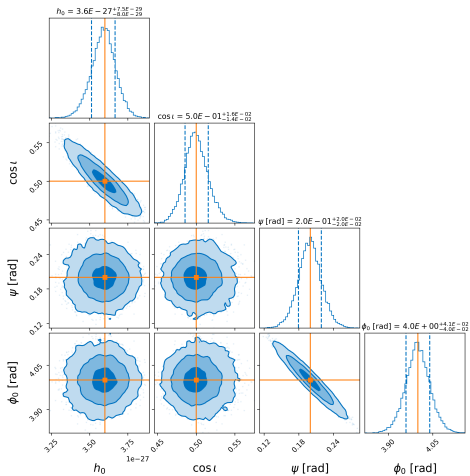
Single signal without noise:



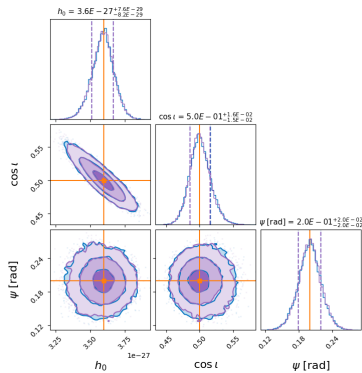
Full likelihood

Tests I

Single signal without noise:



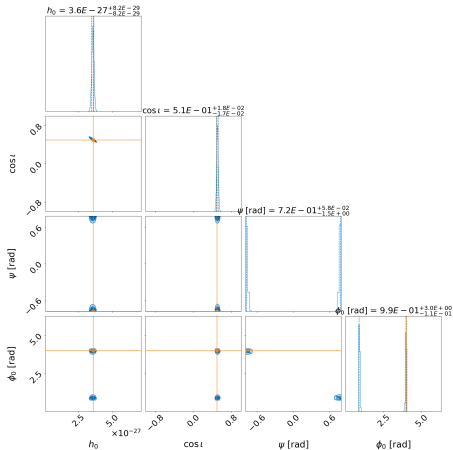
Full likelihood



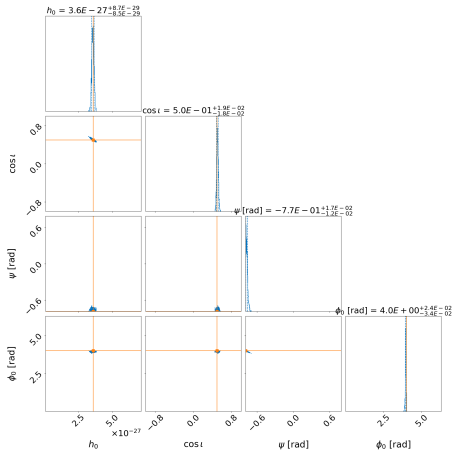
Marginalized likelihood

Tests I: Dynesty vs BilbyMCMC

Recovery of **multi-modal posterior** (signal with $\psi = -\pi/4$):



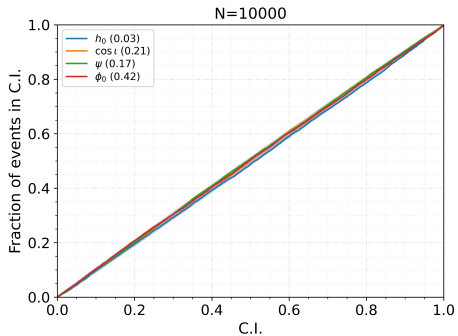
Dynesty



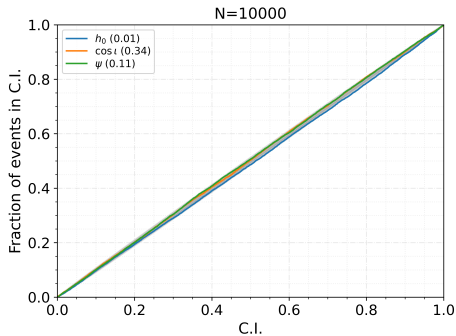
BilbyMCMC

Tests II

PP plots in Gaussian noise: ideally, $x\%$ of the total number of injections should fall in the $x\%$ credible interval



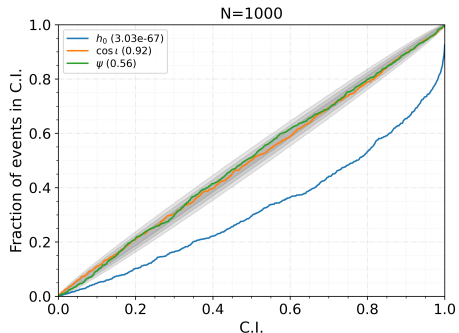
Full likelihood



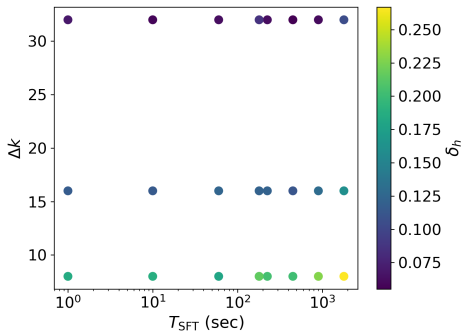
Marginalized likelihood

Tests IIb: biases

The \mathcal{F} -statistic calculation has some assumptions/simplifications that can lead to biases only visible for high SNR signals



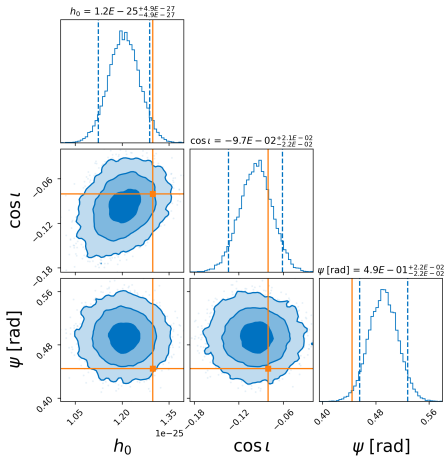
Bias in h_0 for high SNR signals



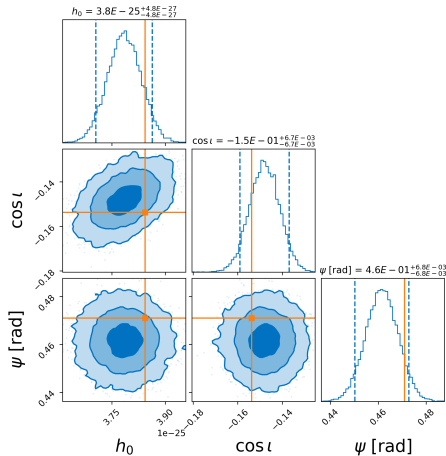
Bias as a function of some \mathcal{F} -statistic calculation-related parameters

Tests III

Hardware injections in real data (also analyzed with the Time-domain heterodyning method for cross-check):



Hardware injection 3



Hardware injection 6

Outline

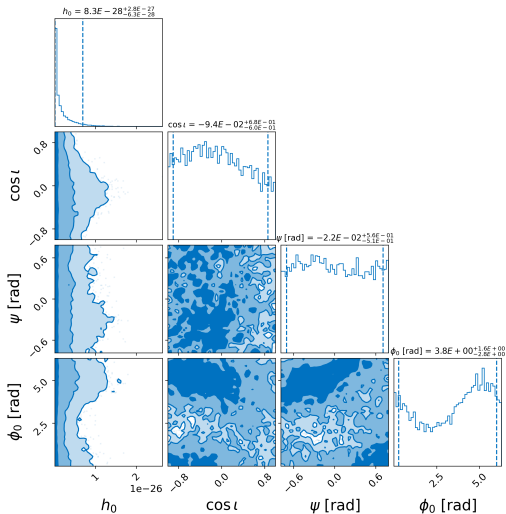
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Search for PSR J1526-2744

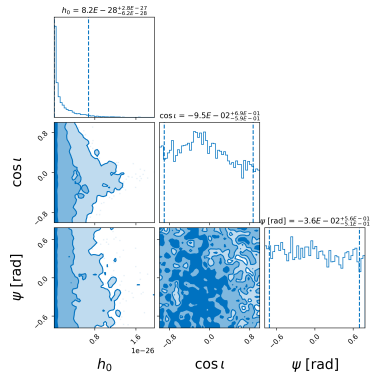
- Discovered in a **joint survey by TRAPUM and FERMI-LAT**, distance of 1.3 kpc and $f_0 = 803.5$ Hz
- Timing solution is derived from 13 years of FERMI-LAT data that overlap with the Advanced LIGO observation runs
- Search done using a **coherent combination of data from the O1, O2, and O3** observation runs of the Advanced LIGO detectors
- **No CW signal is detected**, both with the full and marginalized likelihoods
- 95% upper limit on h_0 is 6.7×10^{-27} , a factor of 9.2 larger than the spin-down upper limit of the pulsar
- 95% upper limit on ϵ is 1.3×10^{-8} assuming a moment of inertia of 10^{38} kg·m²

Posterior distributions for PSR J1526-2744

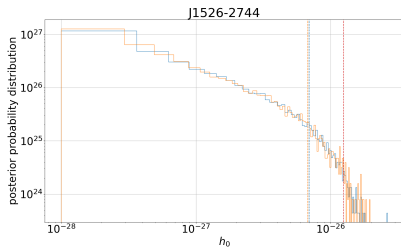
Full likelihood:



Marginalized likelihood:

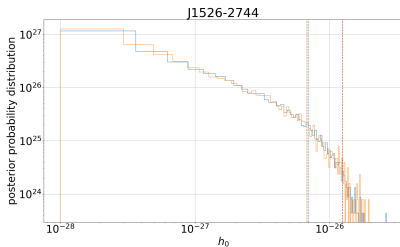


Comparison with frequentist method (in red):

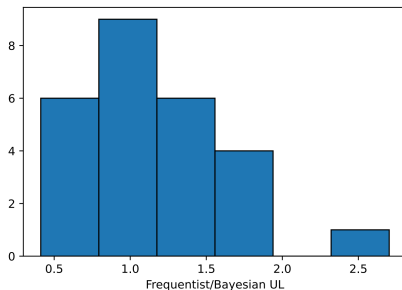


Comparison with frequentist upper limits

Comparison with frequentist method (in red):



Ratio of ULs for some other searched pulsars:



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Future work

- Test limitations of the \mathcal{F} -statistic framework: Gaussian noise assumption, uncertainties of noise floor, Dirichlet kernel and finite SFT length, ...
- Uncertainty on phase-evolution parameters
- More complex signal models, such as signals with more than one frequency or more complicated frequency evolutions
- Characterize usage of Bayesian evidence instead of results from posterior distributions
- Code efficiency comparison to Glasgow pipeline
- Apply framework to more pulsars!

Summary

- A new Bayesian parameter estimation method, based on the highly optimized \mathcal{F} -statistic framework
- Tested with simulated signals, PP plots, and hardware injections
- First application on PSR J1526-2744, no detection!
- Timing solutions from new pulsars are very important

Source material

- A. Ashok et al., PRD 109, 104002 (2024)

Acknowledgements

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