Search for Gravitational Waves from Individual Supermassive Black Hole Binaries in MeerTime Data

> Beatrice Eleonora Moreschi Second year PhD student at University of Milano-Bicocca

Collaborators: Prof. Alberto Sesana, Dr. Golam M. Shaifullah, Dr. Aurélien T. Chalumeau, Dr. Mikel Falxa.

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INAF ISTITUTO NAZIONALE DI ASTROFISICA NATIONAL INSTITUTE FOR ASTROPHYSICS





Sources of gravitational wave in PTA



- Confirm the **existence** of sub-parsec SMBHBs.
- Probe dynamics of SMBHBs.
- Provide test of fundamental physics.



- <u>Deterministic</u>: continuous GW (CGW) generated by inspiralling SMBHBs.
- <u>Stochastic</u>: superimposition of several CGWs or cosmic sources. ²

MeerTime Pulsar Timing Array (MPTA)

100

 10^{-1}

Effelsberg

MeerTime is a survey project of MeerKAT.

64 individual dish antennas with offset Gregorian configuration.

Dataset: ultra precise 88 MSPs (DR2 4.5 yrs).

MeerKAT operates in the range of 856 to 1712 MHz.



The IPTA EPTA Arecibo CPT VLA Parkes NANO CHIME Band-averaged weighted RMS (μ s) InPTA MPT)Gra GMRT MeerKAT

Nançay

Lovell

WSRT

SRT

Build custom models composed of different noises:

- White noise (WN).
- Dispersion measure variation (DM).
- Free chromatic noise.
- Red noise (RN).
- Annual chromatic process.
- Solar Wind (SW).
- Gaussian chromatic bumps.

Systematic issues related to the radiometer. During data collection: folding procedure.

During the travel pulsar-Earth the light crosses ionized material in the interstellar medium. Chromatic noise.

Delay due to **intrinsic instabilities** such as a variation of the pulsar spin. Achromatic noise.

Electron density variations as the line of sight to the pulsar changes during the annual Earth motion around the Sun.

Presence of the Sun between pulsar and Earth. Solar electron density. Constant term + time-varying.

Used to model any deviations from noises.





Optimization of frequency bins in Fourier domain using method in **EPTA II [2023]**.

For RN, DM and chromatic noise.



$$P_{RN}(f) = \frac{A_{RN}^2}{12\pi^2} f_{yr}^{-3} \left(\frac{f}{f_{yr}}\right)^{-\gamma_{RN}}$$

$$P_{DM}(f) = \frac{A_{DM}^2}{12\pi^2} f_{yr}^{-3} \left(\frac{f}{f_{yr}}\right)^{-\gamma_{DM}} \left(\frac{1400MHz}{v}\right)^2$$

$$\left(\mathbf{F}_{j}^{\text{chrom.}}(t_{i}) = \mathbf{F}(t_{i}) * \left(\frac{\nu_{j}}{1.4 \text{ GHz}}\right)^{-\chi}\right)$$

5

1

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$$\mathbf{F}_j^{\text{chrom.}}(t_i) = \mathbf{F}(t_i) * \left(\frac{\nu_j}{1.4 \text{ GHz}}\right)^{-\chi}$$

 $1.4 \, \text{GHz}$

0

Why?

Falxa et al. 2023, Chalumeau et al. 2021. Unmodelled high-frequency noise could conspire for CGW signal.

Build custom models composed of

- different noises:
 - White noise (WN).
 - Dispersion measure variation (DM).
 - Free chromatic noise.
 - Red noise (RN).
 - Annual chromatic process.
 - Solar Wind (SW).
 - Gaussian chromatic bumps.

Optimal model selection between all models (higher **Bayes** factor, not always).

Parameter estimation with PTMCMC sampler of the optimal model.

Check the shape of posterior distributions and if noises are physically motivated (e.g. SW).



• Gaussian chromatic bumps.

and SPNA

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and SPNA



9

CGW search

1.0 (sec)

-0.5

0.5

1.0

2 models for CGW search.

CGW modeled using sine wave template.

 $\begin{aligned} r_a(t) &= r_a^{\mathrm{p}}(t) - r_a^{e}(t) \\ r_a^{e}(t) &= \frac{\mathcal{A}}{\omega} \left\{ (1 + \cos^2 \mathcal{O} F_a^+ [\sin(\omega t + \Phi_0) - \sin \Phi_0] \right] \\ &+ 2\cos\iota F_a^\times [\cos(\omega t + \Phi_0) - \cos \Phi_0] \right\}, \\ r_a^{\mathrm{p}}(t) &= \frac{\mathcal{A}_a}{\omega_a} \left\{ (1 + \cos^2 \iota) F_a^+ [\sin(\omega_a t + \Phi_a + \Phi_0) \\ &- \sin(\Phi_a + \Phi_0)] + 2\cos\iota F_a^\times [\cos(\omega_a t + \Phi_a + \Phi_0) \\ &- \cos(\Phi_a + \Phi_0)] \right\}. \end{aligned}$

Compute the Earth term, no pulsar term.

	CGW parameters	Range
	cosi	[-1, 1]
	$\cos \theta$	[-1, 1]
	$\log_{10} M_{\odot}$	[6.0, 10.0]
	$\log_{10} dL$	[-2.0, 4.0]
3	log ₁₀ f _{CGW}	[-8.097, -6.52]
	ϕ_0	$[0, \pi]$
	ϕ	$[0, 2\pi]$
	Ψ	$[0,\pi]$
!		

1.5

le8

25

Preliminary results

..



CGW+CURN search





Preliminary results of MeerTime **DR2** using **advanced** model selection, **not custom**.

I search CGW in all the sky, blindly.



CGW+CURN search



Survey: DSS colored

CGW search





SMBHs live at the center of the massive galaxies, galaxies merge, SMBHs travel toward the center through:

- Dynamical friction
- Stellar hardening
- Gas torques
- GW emission

Looking for **periodic light curves** in Active Galactic Nuclei (AGN). It traces the SMBHB activity and serve as **electromagnetic counterparts**.



Accretion disk



Looking for **periodic light curves** in Active Galactic Nuclei (AGN).

Build a **catalog** with the best candidates from Survey and literature:

- Catalina Real Time Transient Survey (CRTS, M. J. Graham et al. 2015). Used Sloan Digital Sky Survey (SDSS) spectra to compute masses of SMBHB
- Palomar Transient Factory (PTF)
- Panoramic Survey Telescope & Rapid Response System (Pan-STARRS)
- SDSS J092712.65+294344.0 (M. Dotti et al. 2009)

Object Name	RA	Dec	f_{GW}	$\log(M)$	strain
			[Hz]	$[M_{\odot}]$	h_0
3C66B*	$02 \ 23 \ 11.5$	+42 59 30	6.04E-08	9.08	7.2E-15
HS $1630 + 2355$	$16 \ 33 \ 02.7$	$+23 \ 49 \ 28.8$	1.13E-08	9.86	2.29E-16
SDSS J164452.71+4307	$16 \ 44 \ 52.7$	$+43 \ 07 \ 52.9$	1.16E-08	10.15	4.94E-16
SDSS J114857.33 $+1600$	$11 \ 48 \ 57.4$	$+16 \ 00 \ 22.7$	1.25E-08	9.9	3.02E-16
HS 0926+3608	$09 \ 29 \ 52.1$	+35 54 49.6	1.48E-08	9.95	2.04E-16
SDSS J092911.35+2037	$09 \ 29 \ 11.3$	$+20 \ 37 \ 09.2$	1.30E-08	9.92	2.02E-16
SDSS J133516.17 $+1833$	$13 \ 35 \ 16.1$	$+18 \ 33 \ 41.8$	1.34E-08	9.76	1.91E-16
SDSS J140704.43 $+2735$	$14\ 07\ 04.5$	$+27 \ 35 \ 56.3$	1.48E-08	9.94	1.89E-16
SDSS J134855.27-0321	$13 \ 48 \ 55.3$	$-3 \ 21 \ 41.4$	1.62E-08	9.89	1.78E-16
SDSS J160730.33 $+1449$	$16\ 07\ 30.3$	$+14 \ 49 \ 04.2$	1.34E-08	9.82	1.44E-16
SDSS J131706.19 $+2714$	$13\ 17\ 06.2$	$+27 \ 14 \ 16.7$	1.39E-08	9.92	1.34E-16
SNU J13120+0641	$13\ 12\ 04.7$	$+06 \ 41 \ 07.6$	1.55E-08	9.14	1.33E-16
OJ287	08 54 48.9	+20 06 31	5.82E-09	10.26	1.11E-16

Xin et al. 2021



MPTA has a short dataset, so I consider sources with high frequency GW emission (not long period sources).



Once I finished the SPNA and the catalog is ready:

- I **fix** the candidate's **position** as a first approach.
- Depending on the results we will decide next steps:
 - Can we get any other constraints on parameters? Other check?
 - If useful, we can use other sky surveys for EM counterpart.



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Thank you!





-



1 yr⁻¹

Papers for targeted search:

- Y. J. Chen et al. (2023)
- Ge et al. (2024)
- M. Minev et al. (2024)
- Saade et al. (2024)

Bayes theorem

I used Bayesian techniques, a statistical approach used to analyze data.

In Bayesian framework, **parameters** are treated as random variables. so instead of considering parameters as **fixed** values, a prior probability distribution is assigned to them, which represents our knowledge before observing the data.

"The probability of an event can be estimated from our prior knowledge of conditions that might be related to the event"



Custom SPNA

Details: Chromatic index range [0,14] Annual chromatic index range [0,14]

Then testing for chromatic noise with fixed index -4 (scattering)

$\begin{array}{c ccccccccccccccccccccccccccccccccccc$	$\frac{1}{\sqrt{1+2c}}$		✓ - ✓ -	-	-	
$\begin{array}{c c} M_1 \\ \hline M_2 \\ \hline M_3 \\ \hline M_4 \\ \hline M_5 \\ \hline M_6 \\ \hline M_7 \\ \hline M_8 \\ \hline M_9 \\ \hline M_{10} \\ \hline M_{11} \\ \hline M_{12} \\ \hline M_{13} \\ \hline M_{13} \\ \hline M_{14} \\ \hline M_{15} \\ \hline M_{16} \\ \hline M_{17} \\ \hline M_{18} \\ \hline M_{19} \\ \hline M_{20} \\ \hline M_{21} \\ \hline \end{array}$		✓ ✓ ✓	- ✓ -	-	-	-
$\begin{array}{c c} M_2 \\ M_3 \\ M_4 \\ M_5 \\ M_6 \\ M_7 \\ M_8 \\ M_9 \\ M_{10} \\ M_{11} \\ M_{12} \\ M_{13} \\ M_{14} \\ M_{15} \\ M_{16} \\ M_{17} \\ M_{18} \\ M_{19} \\ M_{20} \\ M_{21} \\ \end{array}$		v	√ -	-	-	1641 a
$\begin{array}{c c} M_3 \\ \hline M_4 \\ \hline M_5 \\ \hline M_6 \\ \hline M_7 \\ \hline M_8 \\ \hline M_9 \\ \hline M_{10} \\ \hline M_{10} \\ \hline M_{11} \\ \hline M_{12} \\ \hline M_{13} \\ \hline M_{13} \\ \hline M_{14} \\ \hline M_{15} \\ \hline M_{16} \\ \hline M_{17} \\ \hline M_{18} \\ \hline M_{19} \\ \hline M_{20} \\ \hline M_{21} \\ \hline \end{array}$	✓ ✓ ✓	v	2			-
$\begin{array}{c c} M_4 \\ M_5 \\ M_6 \\ M_7 \\ M_8 \\ M_9 \\ M_{10} \\ M_{11} \\ M_{12} \\ M_{13} \\ M_{14} \\ M_{15} \\ M_{16} \\ M_{17} \\ M_{18} \\ M_{19} \\ M_{20} \\ M_{21} \\ \end{array}$	√ √	1		~	(- 0	8 1 0
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$ \begin{array}{r} M_{18} \\ M_{19} \\ M_{20} \\ M_{21} \\ \end{array} $	\checkmark	~	~		~	\checkmark
M ₁₉ M ₂₀ M ₂₁	\checkmark	177	-	~	~	\checkmark
M ₂₀ M ₂₁	~	-	~	\checkmark	~	\checkmark
M ₂₁	~	~	-	\checkmark	~	\checkmark
	~	\checkmark	~	\checkmark	\checkmark	\checkmark
M ₂₂	~	-	-	-	-	-
M ₂₃	~	1.00	7	87	~	
<i>M</i> ₂₄		0.75			-	✓
M ₂₅	~	0.25	2	12	~	~

Data prefers the custom model

Why custom?



Using "standard" noise settings: **30** frequency bins for DM and RN. 51 nHz. 21 pulsars.

Time-correlated noise at high frequencies.



Using **custom** noise model for the six best EPTA pulsars (**from Chalumeau et al. 2021**). Posterior distributions not anymore constrained.

SPNA: custom noise model

Optimization of frequency bins in Fourier domain using method in **EPTA II** [2023].

Frequency range for RN, DM and chromatic noise: from 5 to bin corresponding to 1/(2*cadence). 51 Models.

Example of posterior distributions, **code** in progress (thanks Aurélien).







Investigating Gravitational Wave Anisotropy

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Different levels of anisotropy



