Spatio-Temporal Inference Strategies In The Quest For Gravitational Wave Detection With Pulsar Timing Arrays

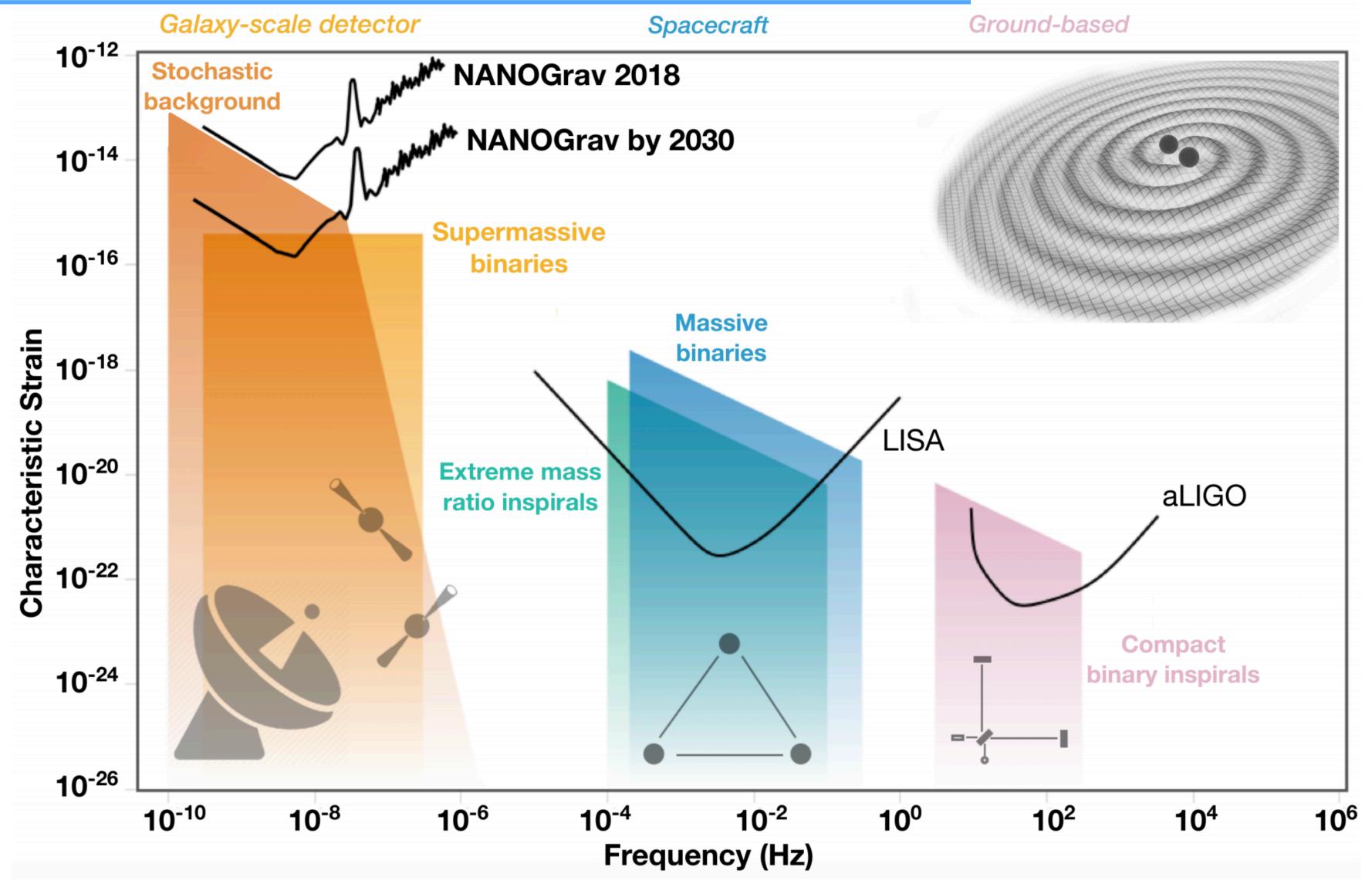


# Stephen Taylor

Vanderbilt University



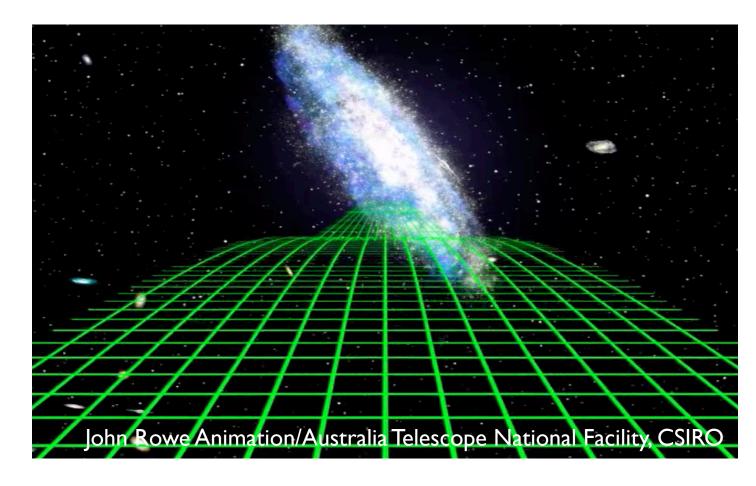


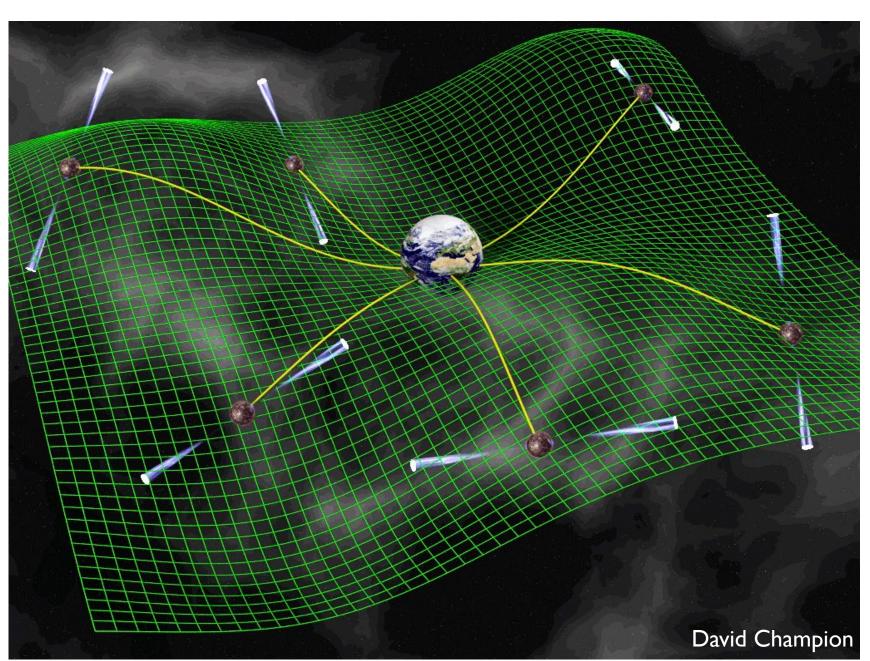


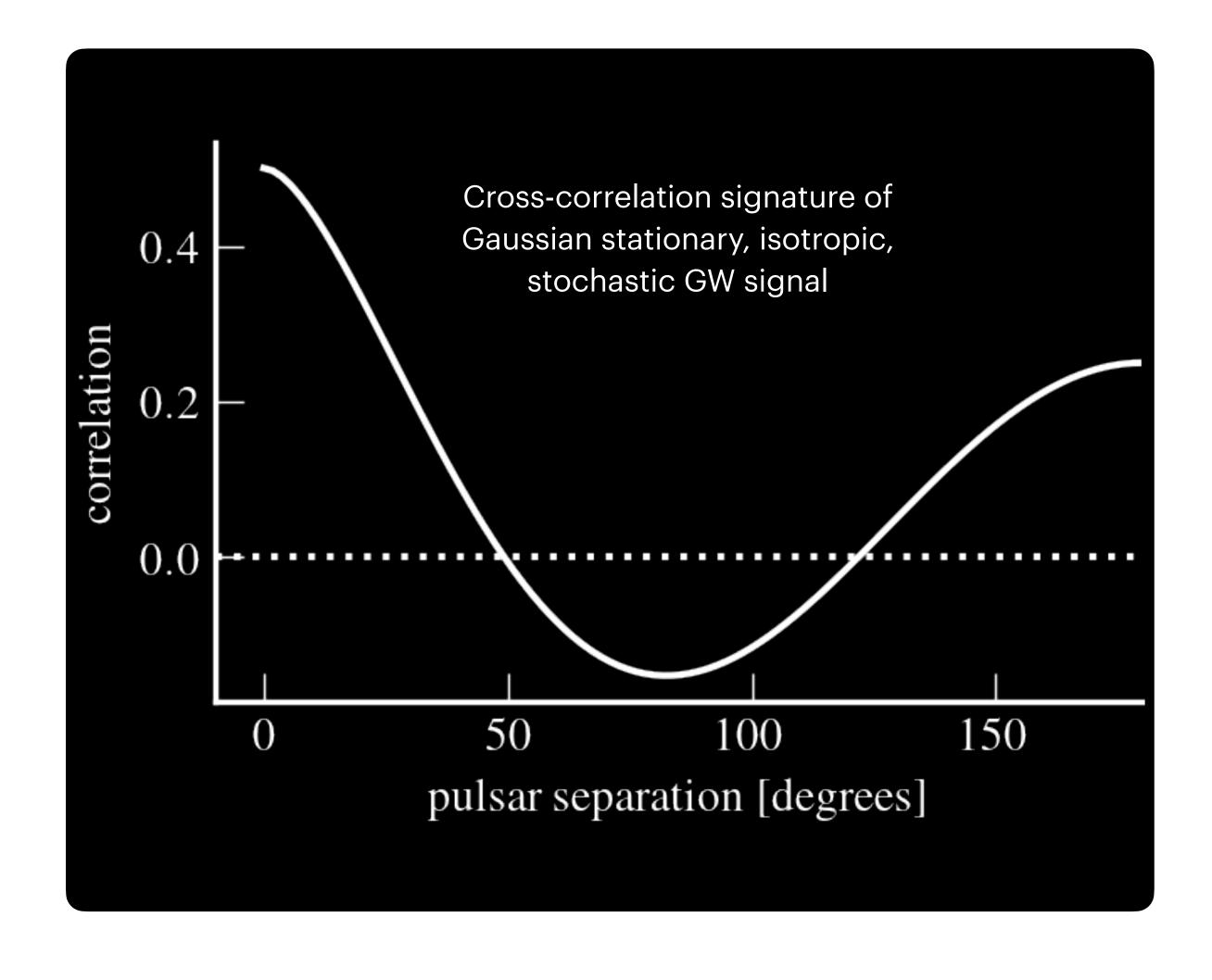
S. Taylor & C. Mingarelli, adapted from gwplotter.org (Moore, Cole, Berry 2014) and based on a figure in Mingarelli & Mingarelli (2018). Illustration of merging black holes adapted from R. Hurt/Caltech-JPL/EPA





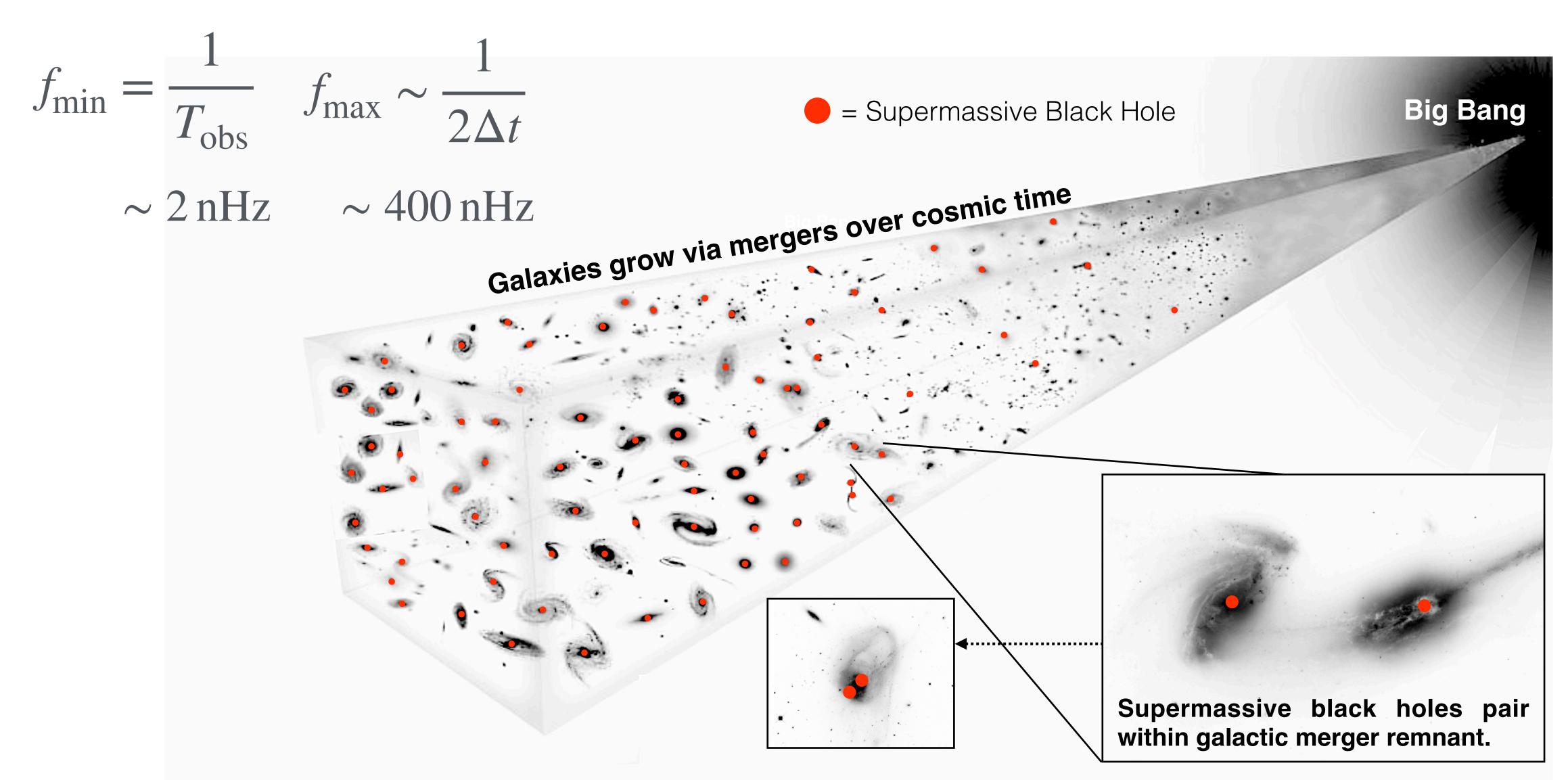






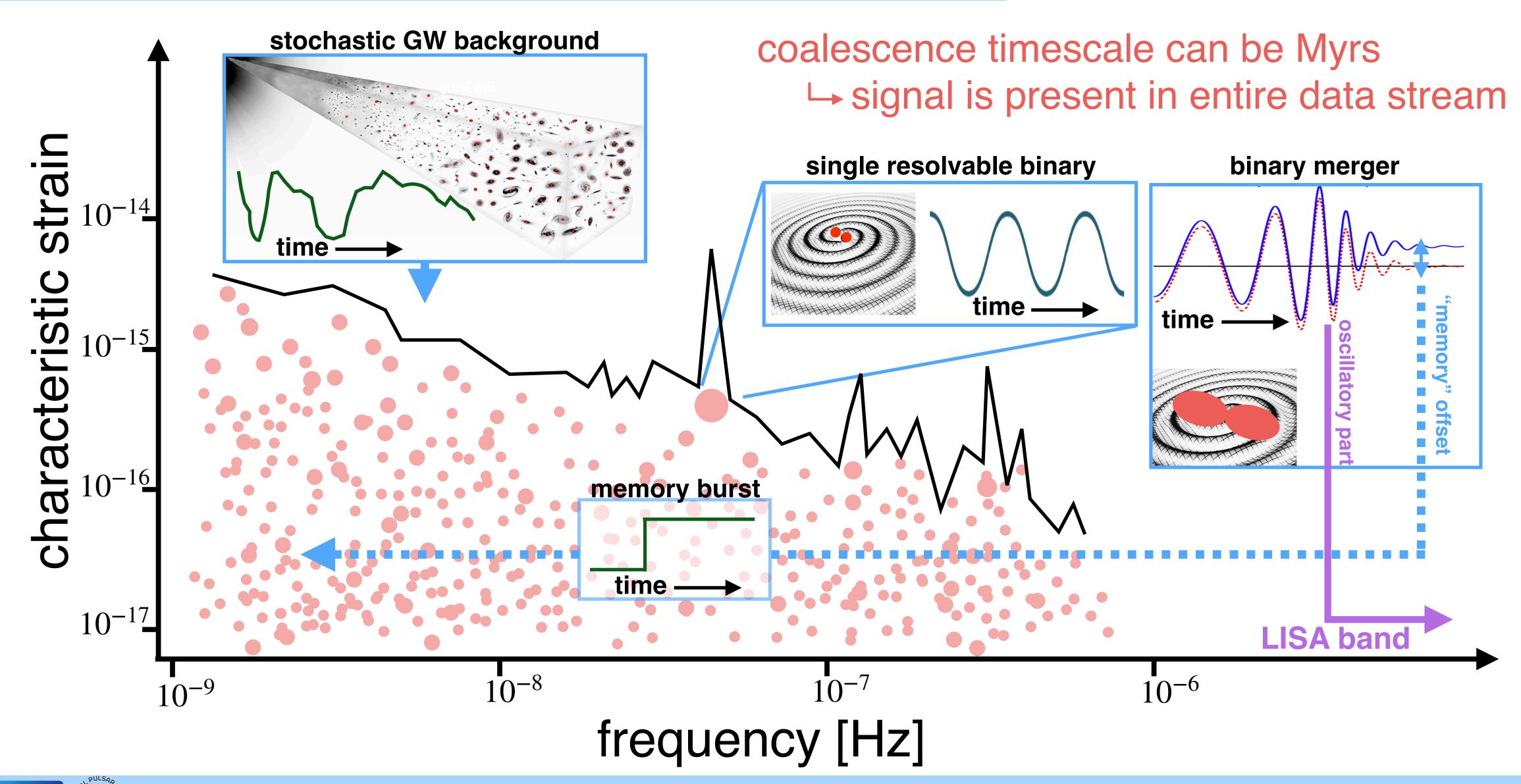












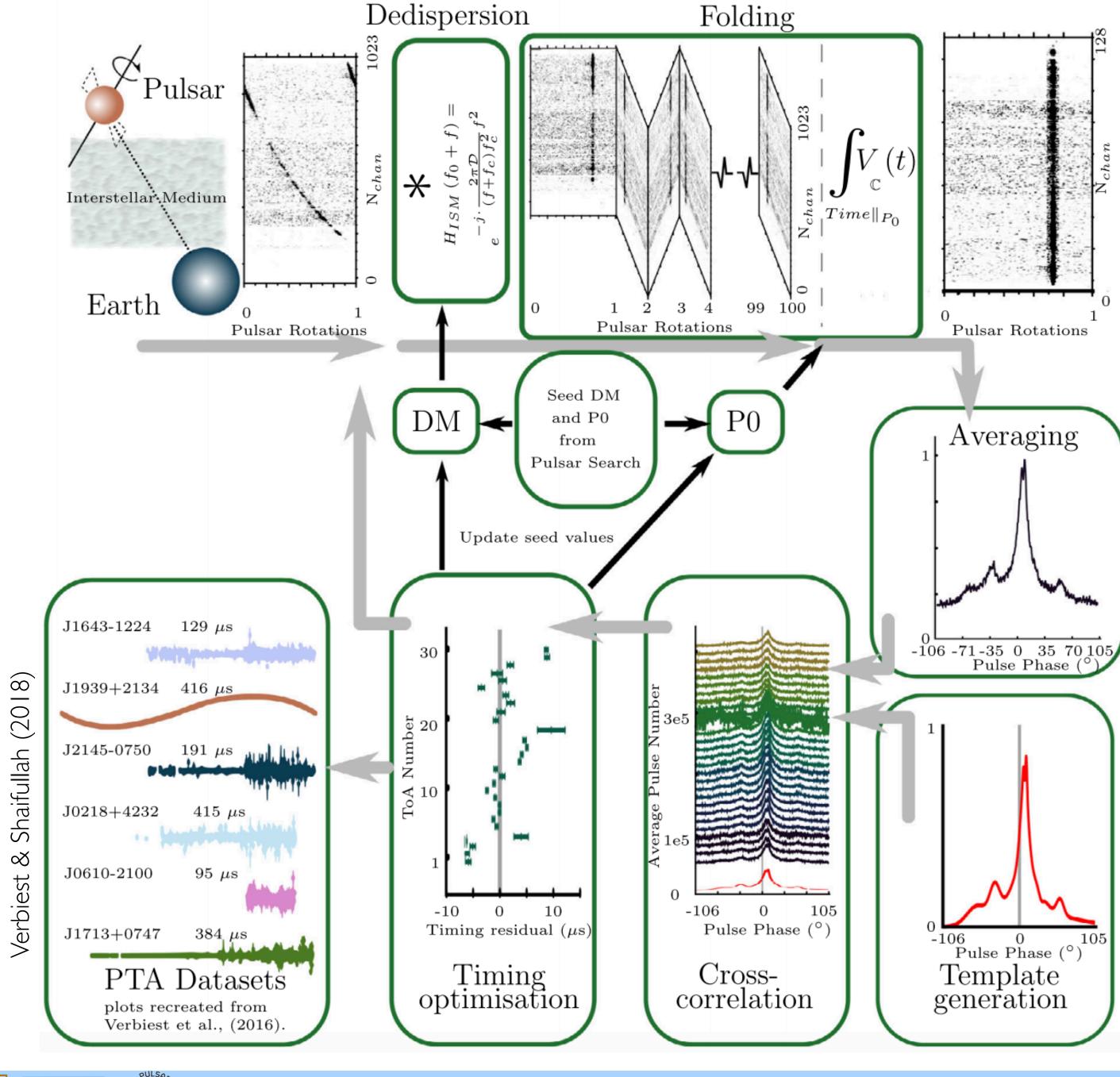












# From pulses to TOAs

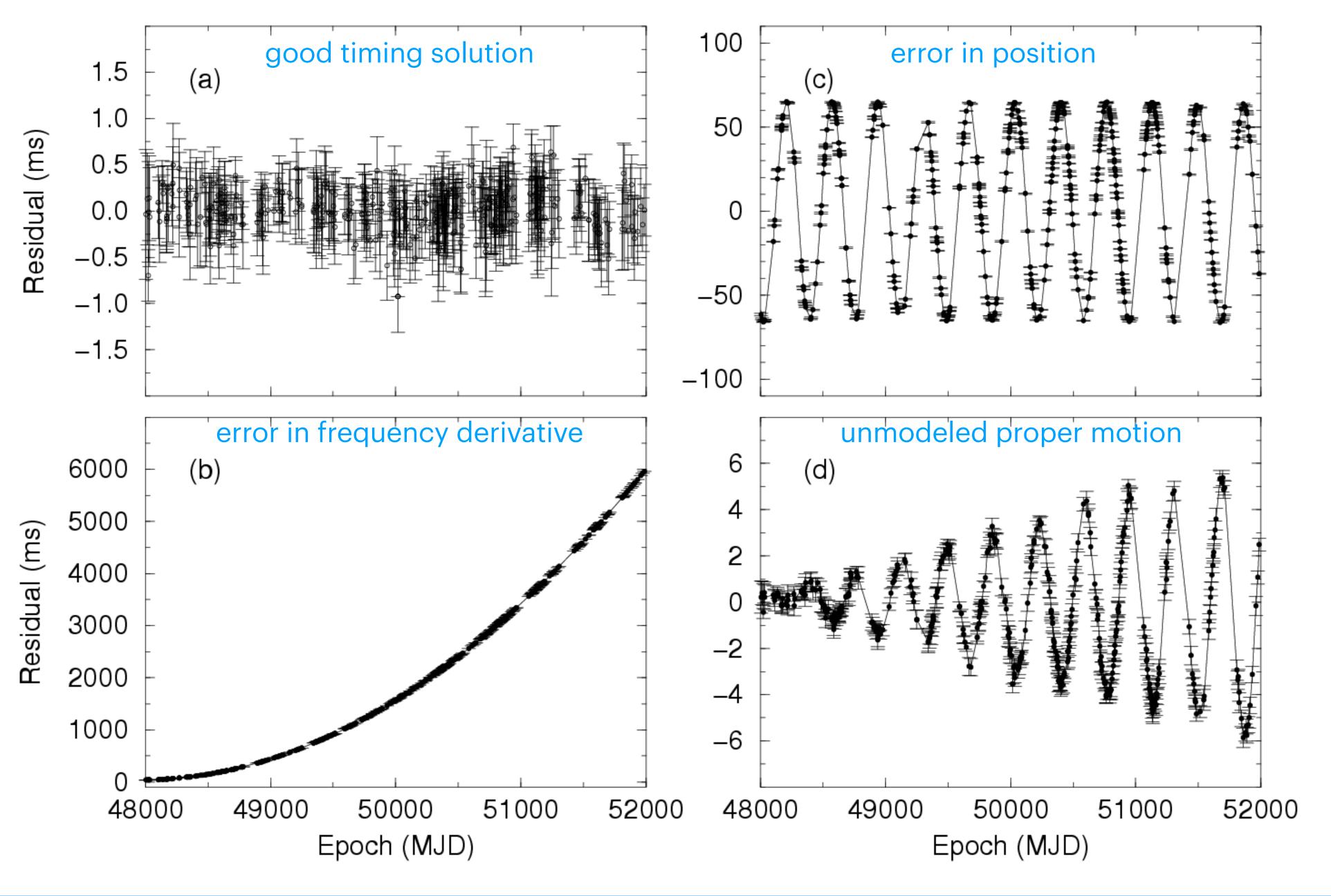
\*TOA = times of arrival







# Creating a timing ephemeris



Lorimer & Kramer (2005)





# Pulsar-timing Data Model

random Gaussian processes

$$\vec{t}_{\text{TOA}} = \vec{t}_{\text{det}} + \vec{t}_{\text{stoch}}$$

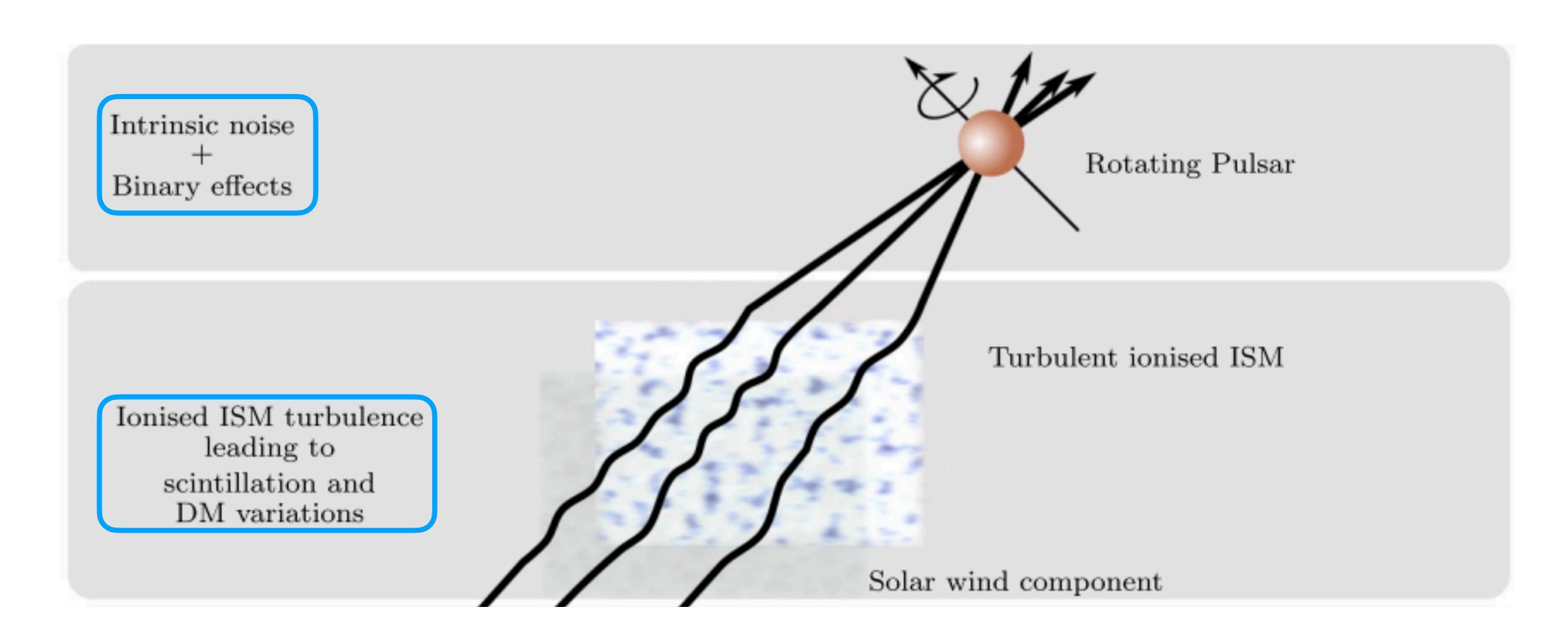
$$\overrightarrow{\delta t} \equiv \overrightarrow{t}_{TOA} - \overrightarrow{t}_{det}(\overrightarrow{\beta}_0) \qquad \longleftarrow \text{Timing residuals}$$

Deterministic	Stochastic
timing ephemeris	per-pulsar achromatic red noise
	per-pulsar white noise
transient noise features	per-pulsar chromatic red noise
single resolvable GW signals	interpulsar-correlated achromatic processes





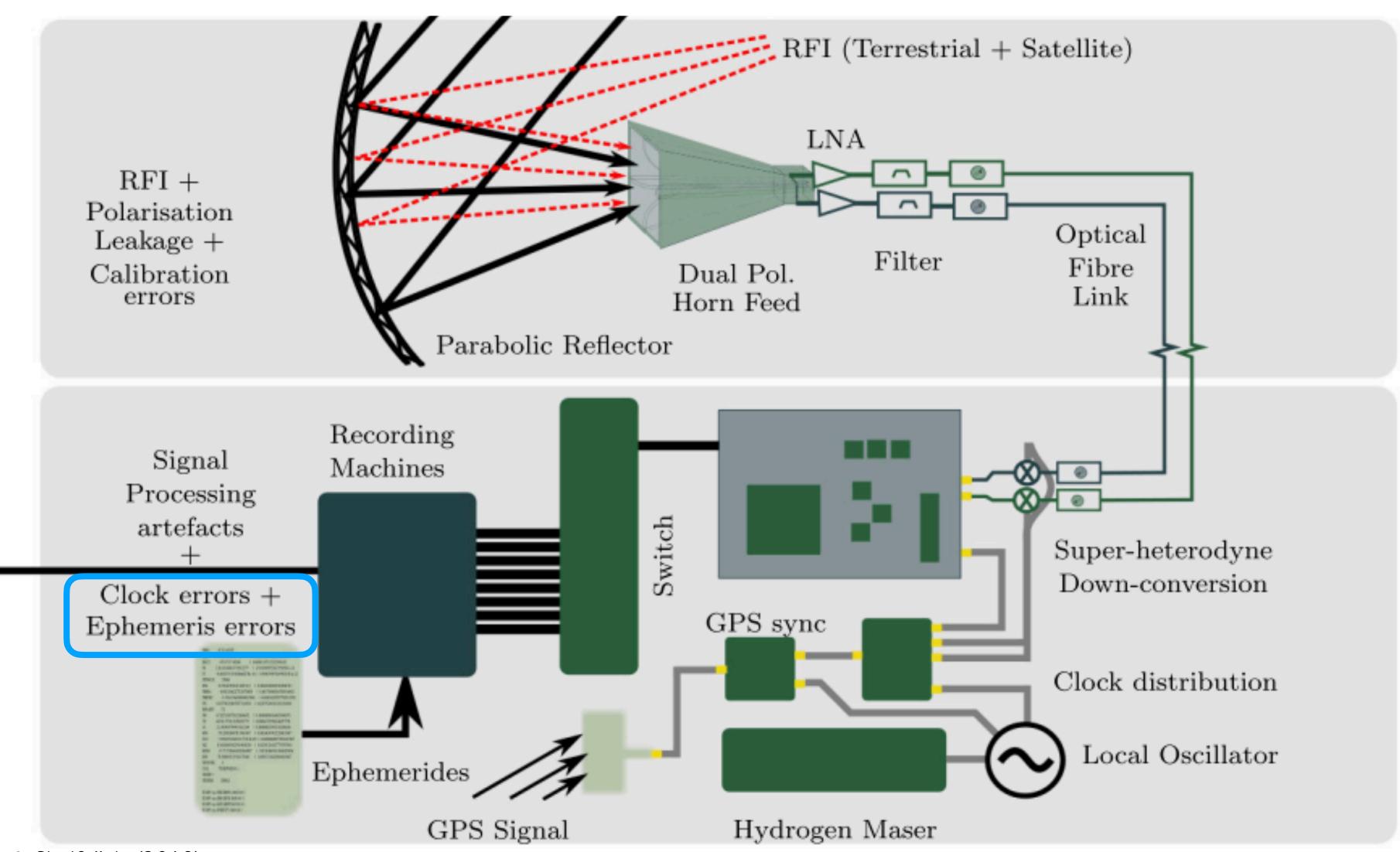
#### Sources of noise



Verbiest & Shaifullah (2018)



#### Sources of noise

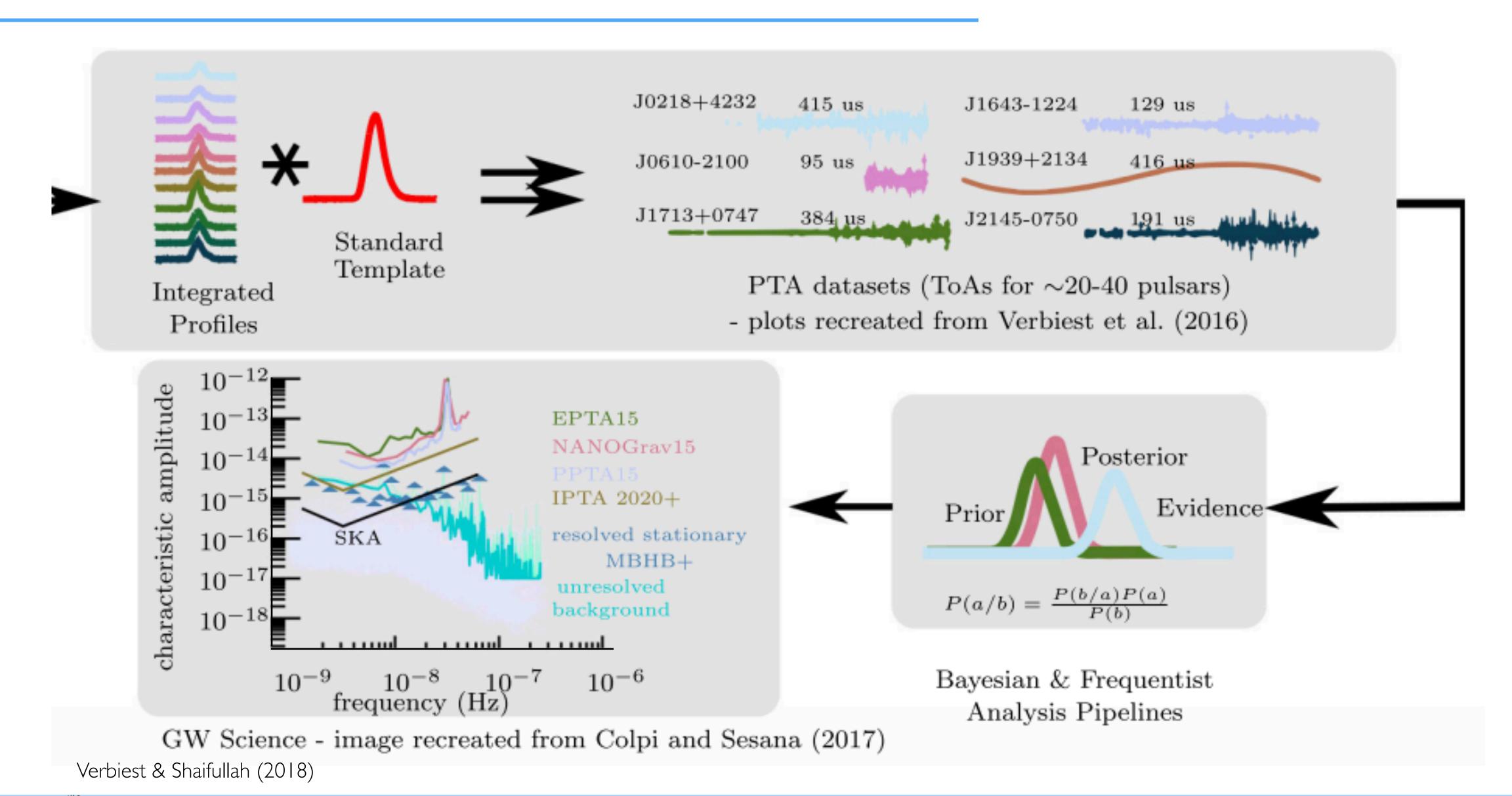


Verbiest & Shaifullah (2018)





## Sources of noise





# Pulsar-timing Data Model

$$\delta t = \delta t_{\rm tm} + \delta t_{\rm white} + \delta t_{\rm red}$$

- Deviations around best-fit of timing ephemeris
- White noise
  - TOA measurement uncertainties
  - Extra unaccounted white-noise from receivers
  - Pulse phase "jitter"

- Intrinsic low-frequency processes
  - Rotational instabilities lead to random walk in phase, period, period-derivative
  - Radio-frequency dependent dispersionmeasure variations
- Spatially-correlated low-frequency processes
  - Stochastic variations in time standards
  - Solar-system ephemeris errors
  - Gravitational-wave background





# Timing Ephemeris

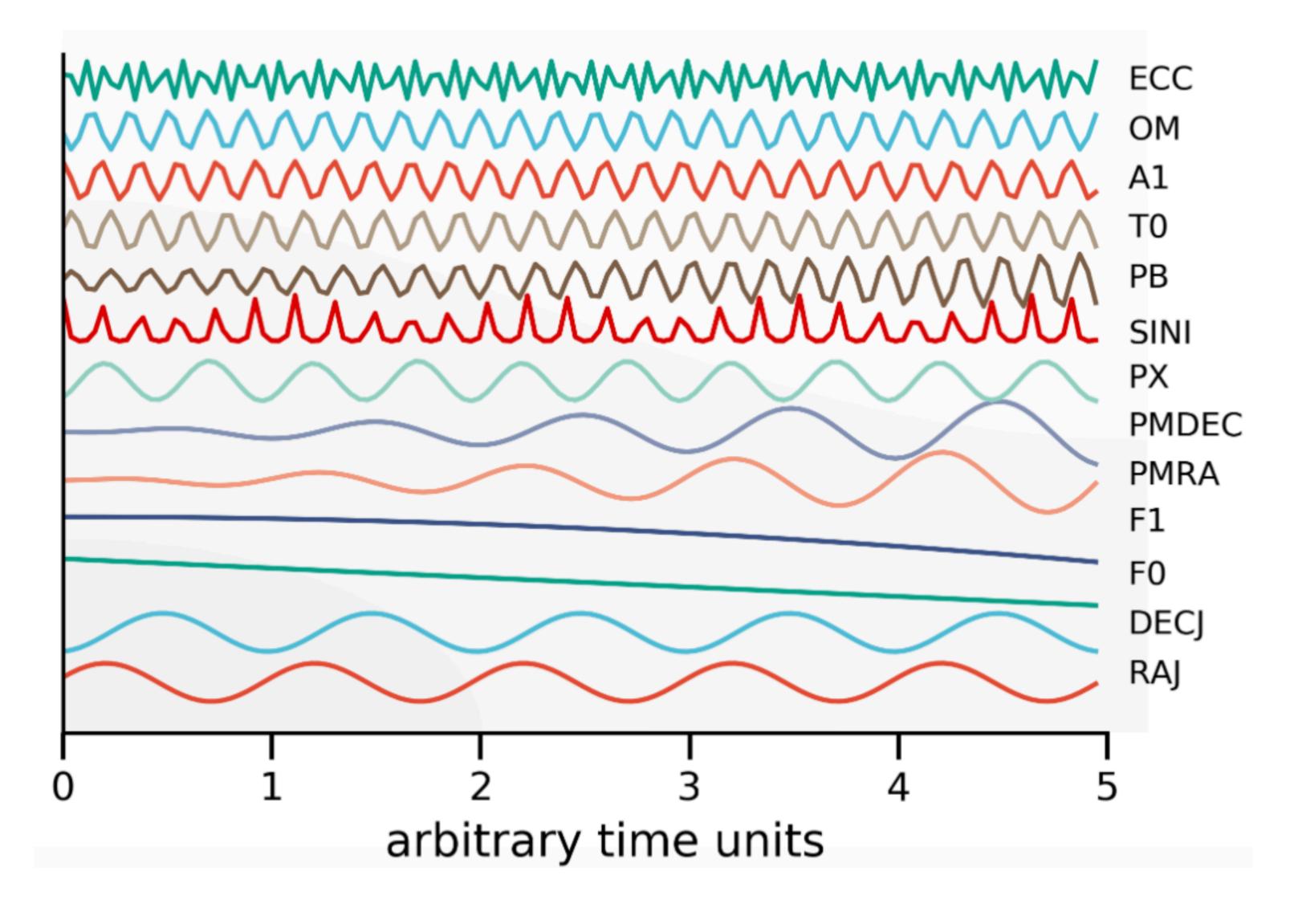
$$t_{\text{det},i}(\overrightarrow{\beta}) = t_{\text{det},i}(\overrightarrow{\beta}_0) + \left[ \sum_{j} \frac{\partial t_{\text{det},i}}{\partial \beta_j} \middle|_{\overrightarrow{\beta}_0} \times (\beta_j - \beta_{0,j}) \right]$$

$$\vec{t}_{\text{det}}(\vec{\beta}) = \vec{t}_{\text{det}}(\vec{\beta}_0) + \mathbf{M}\vec{\epsilon}$$

Timing ephemeris design matrix for linear offsets

# Timing Ephemeris

Temporal behavior of timing ephemeris basis

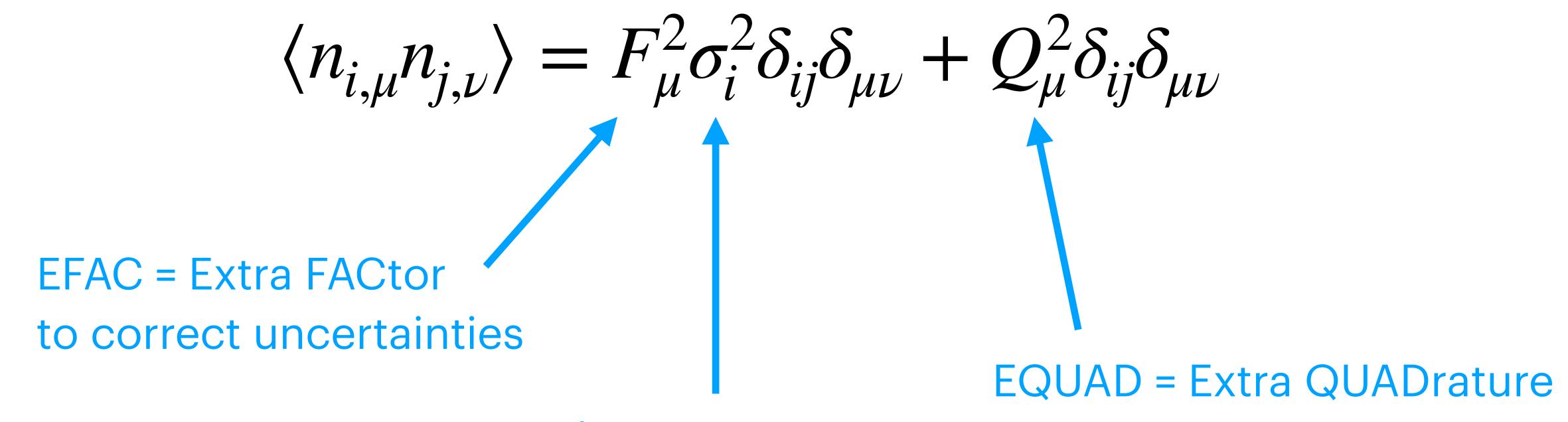






# White Noise (1/2)

- Flat power-spectral density across all sampling frequencies
- No inter-pulsar correlations



"Radiometer noise"—
pulse template fitting
uncertainties

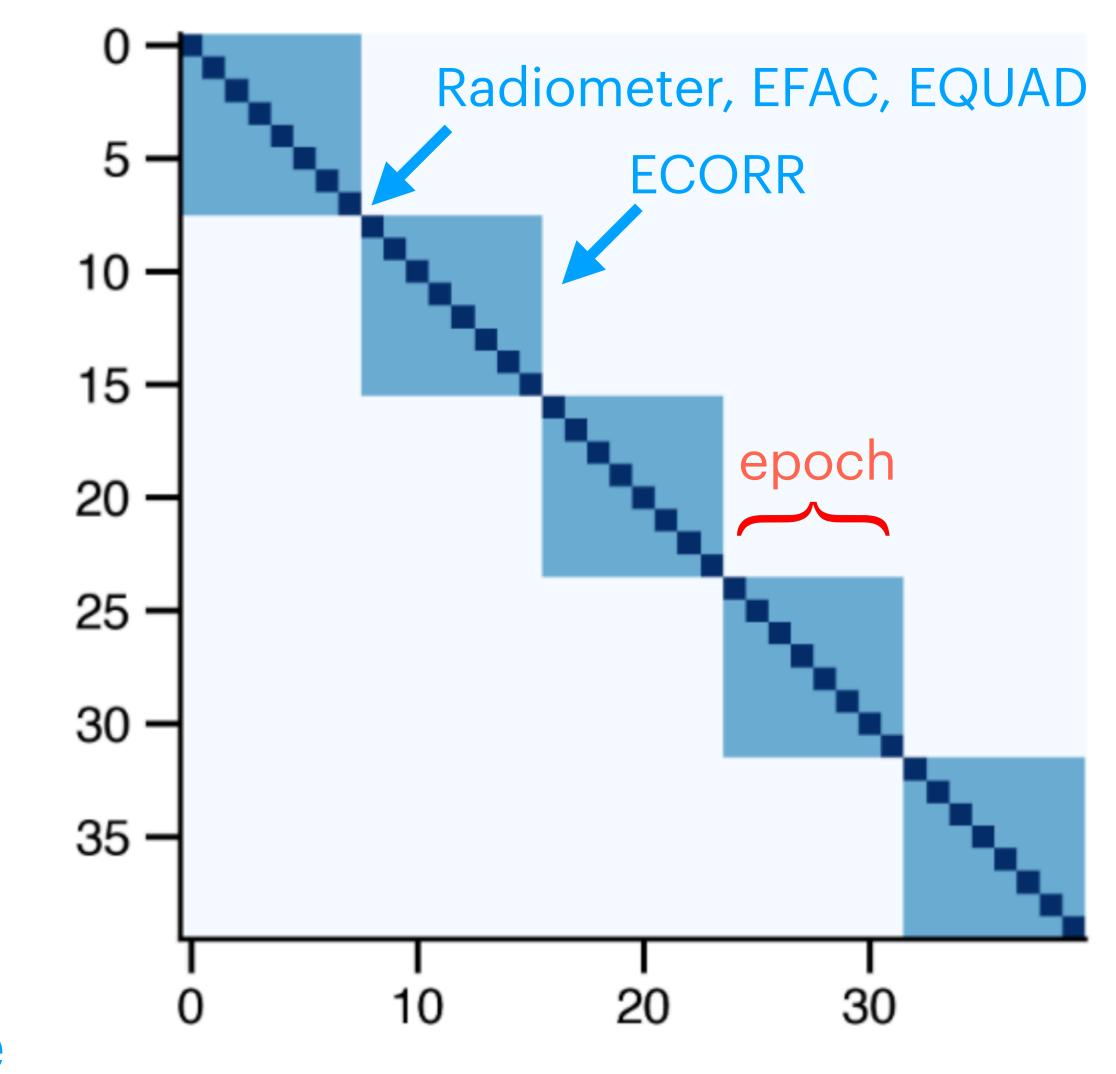


# White Noise (2/2)

- Fitting a template to a finite-pulse folded observation can give "jitter" errors
- Simultaneous observations across many radio sub-bands in an epoch will have correlated jitter errors

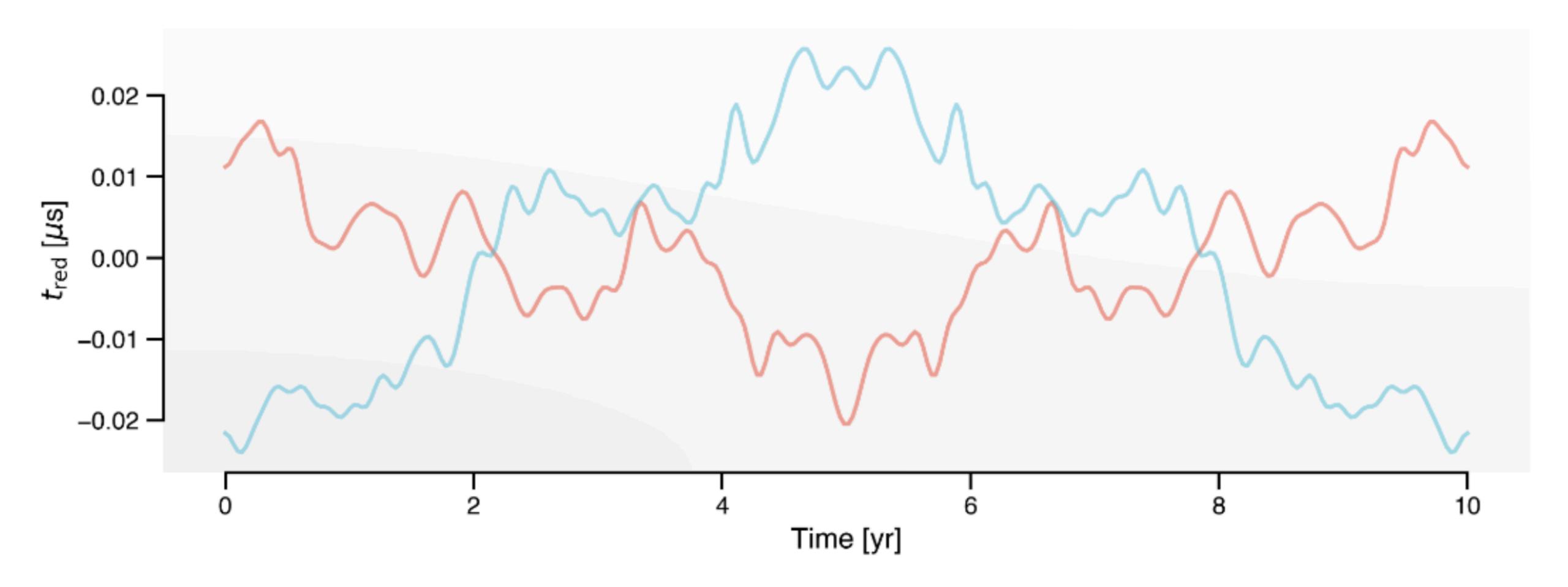
$$\langle n_{i,\mu}^J n_{j,\nu}^J \rangle = J_{\mu}^2 \delta_{e(i)e(j)} \delta_{\mu\nu}$$

ECORR = Extra CORRelated white noise





# Red Processes (1/5)







# Red Processes (2/5)

• Time-domain covariance matrix is large and dense  $\langle \delta t_i \delta t_i \rangle = C(|t_i - t_i|)$ 

$$\langle \delta t_i \delta t_j \rangle = C(|t_i - t_j|)$$

- But we only care abut the lowest frequencies
- Use a rank-reduced formalism for covariance

$$\overrightarrow{\delta t}_{\mathrm{red}} = \mathbf{F} \overrightarrow{a}$$

$$\langle \overrightarrow{\delta t}_{\text{red}} \overrightarrow{\delta t}_{\text{red}}^{\text{T}} \rangle = \mathbf{F} \langle \overrightarrow{a} \overrightarrow{a}^{\text{T}} \rangle \mathbf{F}^{\text{T}}$$
$$C = \mathbf{F} \phi \mathbf{F}^{\text{T}}$$





# Red Processes (3/5)

$$\overrightarrow{\delta t}_{\mathrm{red}} = \overrightarrow{Fa}$$

Fourier design matrix over small number of modes

$$\mathbf{F} = \begin{pmatrix} \sin(2\pi t_1/T) \cos(2\pi t_1/T) \cdots \sin(2\pi N_f t_1/T) \cos(2\pi N_f t_1/T) \\ \sin(2\pi t_2/T) \cos(2\pi t_2/T) \cdots \sin(2\pi N_f t_2/T) \cos(2\pi N_f t_2/T) \\ \vdots & \vdots & \vdots & \vdots \\ \sin(2\pi t_N/T) \cos(2\pi t_N/T) \cdots \sin(2\pi N_f t_N/T) \cos(2\pi N_f t_N/T) \end{pmatrix}$$



## Red Processes (4/5)

$$\delta t_{\text{red}} = \mathbf{F} \overrightarrow{a}$$
Fourier coefficients
$$p(\overrightarrow{a} | \overrightarrow{\eta}) = \frac{\exp\left(-\frac{1}{2}\overrightarrow{a}^{\text{T}}\phi(\overrightarrow{\eta})^{-1}\overrightarrow{a}\right)}{\sqrt{\det(2\pi\phi(\overrightarrow{\eta}))}}$$

$$[\phi]_{(ak)(bj)} = \Gamma_{ab}\rho_k\delta_{kj} + \kappa_{ak}\delta_{kj}\delta_{ab}$$
Overlap Reduction Function
GWB PSD
Intrinsic red-noise PSD





# Red Processes (5/5)

GWB PSD

$$\rho(f) = S(f)\Delta f = \frac{h_c(f)^2}{12\pi^2 f^3} \frac{1}{T}$$

power laws

per frequency

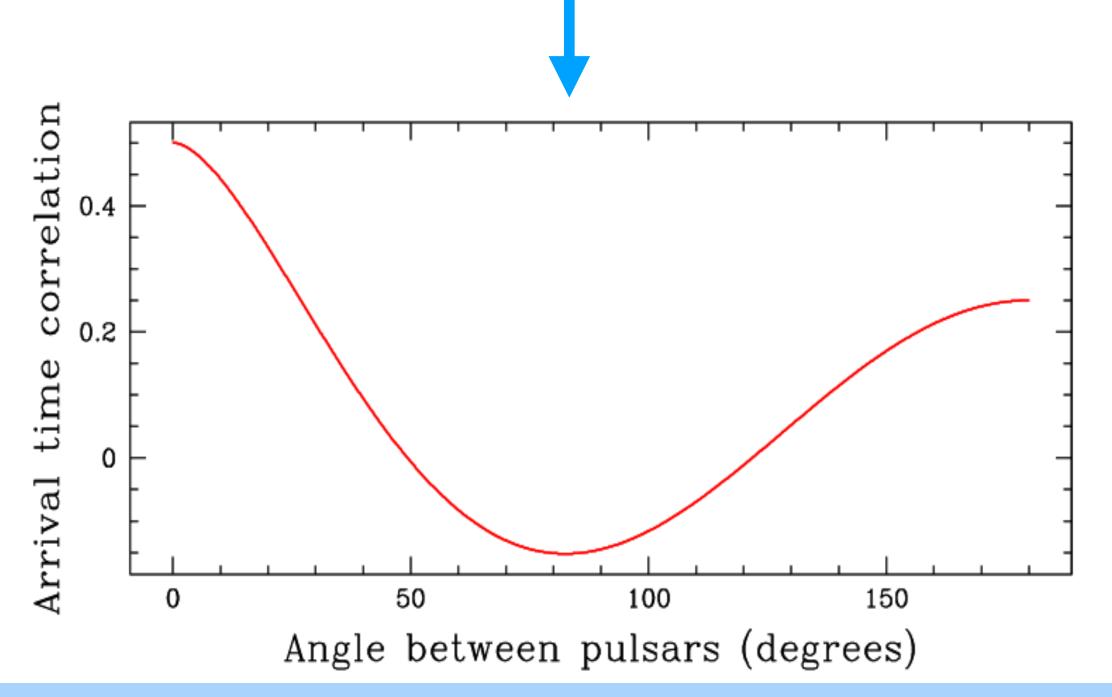
GP emulators

GWB ORF

$$\Gamma_{ab} \propto (1 + \delta_{ab}) \int_{S^2} d^2 \hat{\Omega} P(\hat{\Omega}) \left[ F_a^{\dagger}(\hat{\Omega}) F_b^{\dagger}(\hat{\Omega}) + F_a^{\times}(\hat{\Omega}) F_b^{\times}(\hat{\Omega}) \right]$$

PTA overlap reduction function for Gaussian stationary, isotropic stochastic GWB

"Hellings & Downs Curve" (1983)



$$\overrightarrow{\delta t} = \overrightarrow{M} \overrightarrow{\epsilon} + F \overrightarrow{a} + U \overrightarrow{j} + \overrightarrow{n}$$

$$\overrightarrow{small linear perturbations}$$

$$\overrightarrow{around best-fit timing solution}$$

$$\overrightarrow{low-frequency processes}$$

$$\overrightarrow{in Fourier basis}$$

$$\overrightarrow{jitter}$$

$$\overrightarrow{white noise}$$

$$[M] = N_{\text{TOA}} \times N_{\text{tm}}$$
$$[\vec{\epsilon}] = N_{\text{tm}}$$

"M" is matrix of TOA derivatives wrt timing-model parameters

~ few tens

$$[F] = N_{\text{TOA}} \times 2N_{\text{freqs}}$$
  
 $[\overrightarrow{a}] = 2N_{\text{freqs}}$ 

"F" has columns of sines and cosines for each frequency

~ few tens

$$[U] = N_{\text{TOA}} \times N_{\text{epochs}}$$
$$[j] = N_{\text{epochs}}$$

"U" has block diagonal structure, with ones filling each block

~ couple of hundred

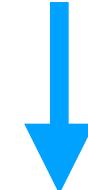
Lentati et al. (inc Taylor) (2013) van Haasteren & Vallisneri (2014a,b)







Start with Gaussian white noise likelihood 
$$p(\overrightarrow{n}) = \frac{\exp\left(-\frac{1}{2}\overrightarrow{n}^{\mathsf{T}}N^{-1}\overrightarrow{n}\right)}{\sqrt{\det(2\pi N)}}$$



$$p(\overrightarrow{\delta t} \mid \overrightarrow{\epsilon}, \overrightarrow{a}, \overrightarrow{j}) = \frac{\exp\left[-\frac{1}{2}\left(\overrightarrow{\delta t} - M\overrightarrow{\epsilon} - F\overrightarrow{a} - U\overrightarrow{j}\right)^{\mathrm{T}}N^{-1}\left(\overrightarrow{\delta t} - M\overrightarrow{\epsilon} - F\overrightarrow{a} - U\overrightarrow{j}\right)\right]}{\sqrt{\det(2\pi N)}}$$

$$p(\overrightarrow{\delta t} \mid \overrightarrow{b}) = \frac{\exp\left[-\frac{1}{2}\left(\overrightarrow{\delta t} - T\overrightarrow{b}\right)^{\mathrm{T}} N^{-1}\left(\overrightarrow{\delta t} - T\overrightarrow{b}\right)\right]}{\sqrt{\det(2\pi N)}}$$

$$\overrightarrow{Tb} = \overrightarrow{M} \overrightarrow{\epsilon} + F \overrightarrow{a} + U \overrightarrow{j}$$

$$\mathbf{b} = \begin{bmatrix} \epsilon \\ \mathbf{a} \\ \mathbf{j} \end{bmatrix} \qquad T = [M \quad F \quad U]$$







#### But we're describing all stochastic terms as random Gaussian processes...

$$p(\overrightarrow{b} \mid \overrightarrow{\eta}) = \frac{\exp\left(-\frac{1}{2}\overrightarrow{b}^{\mathrm{T}}\mathbf{B}^{-1}\overrightarrow{b}\right)}{\sqrt{\det(2\pi\mathbf{B})}}$$

$$m{B} = egin{pmatrix} m{\infty} & m{0} \ m{0} & m{\phi} \end{pmatrix}$$

$$p(\overrightarrow{\eta}, \overrightarrow{b} \mid \overrightarrow{\delta t}) \propto p(\overrightarrow{\delta t} \mid \overrightarrow{b}) p(\overrightarrow{b} \mid \overrightarrow{\eta}) p(\overrightarrow{\eta})$$

hierarchical modelling

$$p(\overrightarrow{\eta} | \overrightarrow{\delta t}) = \int p(\overrightarrow{\eta} | \overrightarrow{\delta t}) d\overrightarrow{b}$$

(analytically!) marginalize over coefficients

$$\Rightarrow p(\overrightarrow{\eta} \mid \overrightarrow{\delta t}) \propto \frac{\exp\left(-\frac{1}{2}\overrightarrow{\delta t}^{\mathrm{T}}\mathbf{C}^{-1}\overrightarrow{\delta t}\right)}{\sqrt{\det(2\pi\mathbf{C})}}p(\overrightarrow{\eta})$$

 $C = N + TBT^{\mathrm{T}}$ 

$$C = N + TBT^{\mathrm{T}}$$

what are we actually doing here?

$$[TBT^{T}]_{(ab),\tau} = \sum_{k}^{N_f} [\phi]_{ab} \cos(2\pi k\tau/T)$$

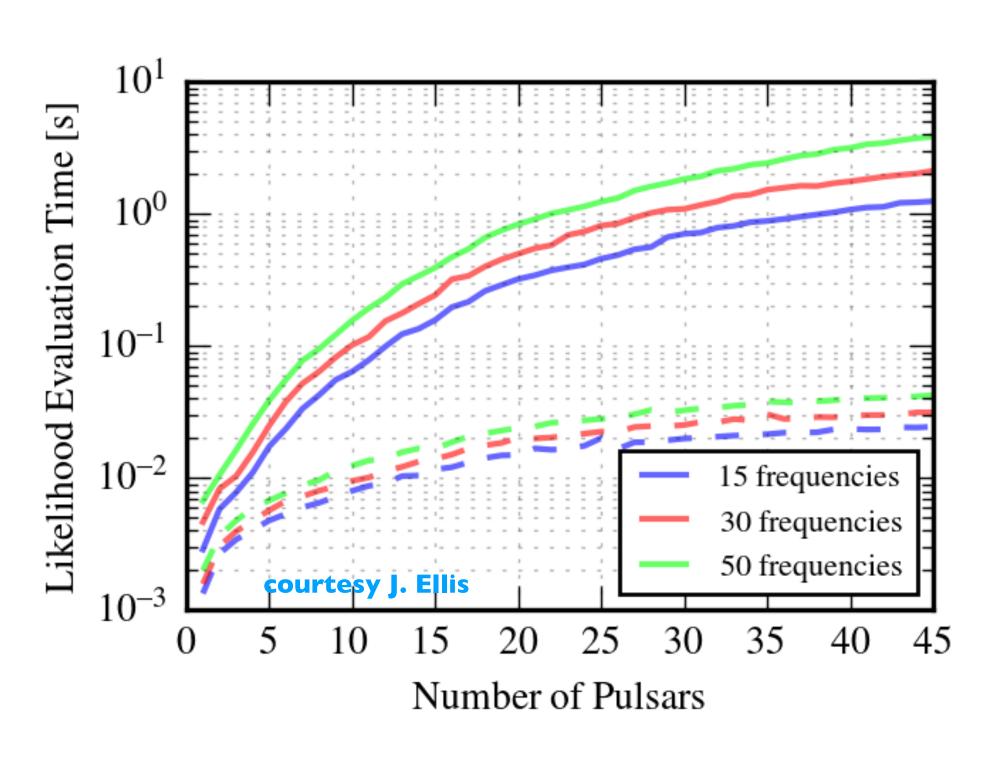
this is just the Wiener-Khinchin theorem!

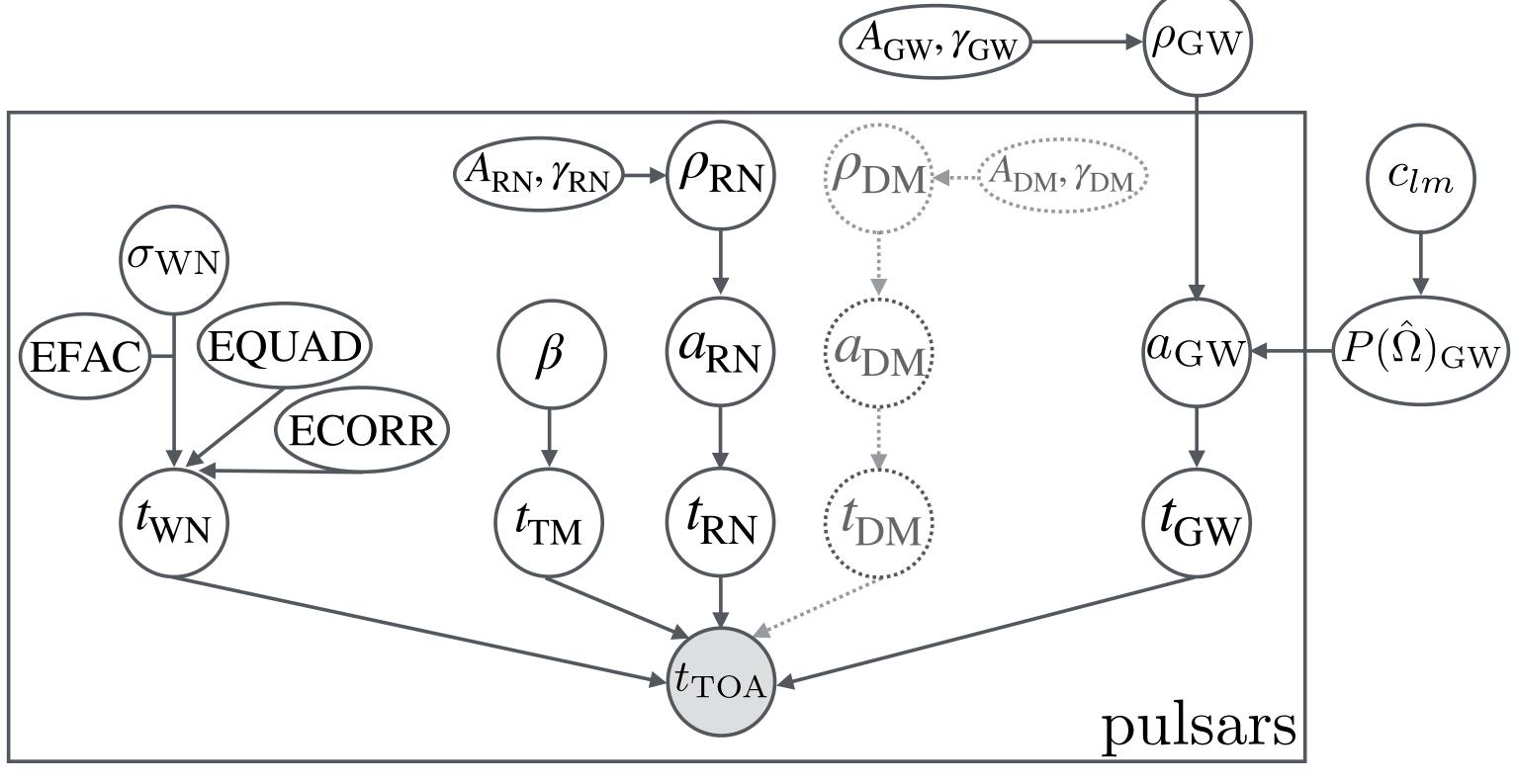
#### Woodbury lemma

$$C^{-1} = (N^{-1} + TBT^{T})^{-1}$$

$$= N^{-1} - N^{-1}T(B^{-1} + T^{T}N^{-1}T)^{-1}T^{T}N^{-1}$$

Much easier and faster than  $N_{\mathrm{TOA}} \times N_{\mathrm{TOA}}$  inversion





Without inter-pulsar correlations [~ tens of ms]

With inter-pulsar correlations [~few seconds]

The PTA Bayesian Network





#### The NANOGrav 12.5-year Data Set: Search For An Isotropic Stochastic Gravitational-Wave Background

Zaven Arzoumanian,¹ Paul T. Baker,² Harsha Blumer,³,⁴ Bence Bécsy,⁵ Adam Brazier,⁶ Paul R. Brook,³,⁴ Sarah Burke-Spolaor,³,⁴,⁵ Shami Chatterjee,⁶ Siyuan Chen,<sup>8,9,10</sup> James M. Cordes,⁶ Neil J. Cornish,⁶ Fronefield Crawford,¹¹ H. Thankful Cromartie,¹² Megan E. DeCesar,¹³,¹,¹⁴,∗ Paul B. Demorest,¹⁵ Timothy Dolch,¹⁶ Justin A. Ellis,¹⁶ Elizabeth C. Ferrara,¹ð William Fiore,³,⁴ Emmanuel Fonseca,¹⁰ Nathan Garver-Daniels,³,⁴ Peter A. Gentile,³,⁴ Deborah C. Good,²⁰ Jeffrey S. Hazboun,²¹,∗ A. Miguel Holgado,²² Kristina Islo,²³ Ross J. Jennings,⁶ Megan L. Jones,²³ Andrew R. Kaiser,³,⁴ David L. Kaplan,²³ Luke Zoltan Kelley,²⁴ Joey Shapiro Key,²¹ Nima Laal,²⁵ Michael T. Lam,²⁶,²ጾ T. Joseph W. Lazio,²⁵ Duncan R. Lorimer,³,⁴ Jing Luo,²⁰ Ryan S. Lynch,³⁰ Dustin R. Madison,³,⁴ \* Maura A. McLaughlin,³,⁴ Chiara M. F. Mingarelli,³¹,³² Cherry Ng,³³ David J. Nice,¹³ Timothy T. Pennucci,³⁴,³⁵,∗ Nihan S. Pol,³,⁴ Scott M. Ransom,³⁴ Paul S. Ray,³⁶ Brent J. Shapiro-Albert,³,⁴ Xavier Siemens,²⁵,²³ Joseph Simon,²²,³ Renée Spiewak,³⁵ Ingrid H. Stairs,²⁰ Daniel R. Stinebring,³⁰ Kevin Stovall,¹⁵ Jerry P. Sun,²⁵ Joseph K. Swiggum,¹³,∗ Stephen R. Taylor,⁴⁰ Jacob E. Turner,³,⁴ Michele Vallisneri,²⁵ Sarah J. Vigeland,²³ Caitlin A. Witt,³,⁴

THE NANOGRAV COLLABORATION

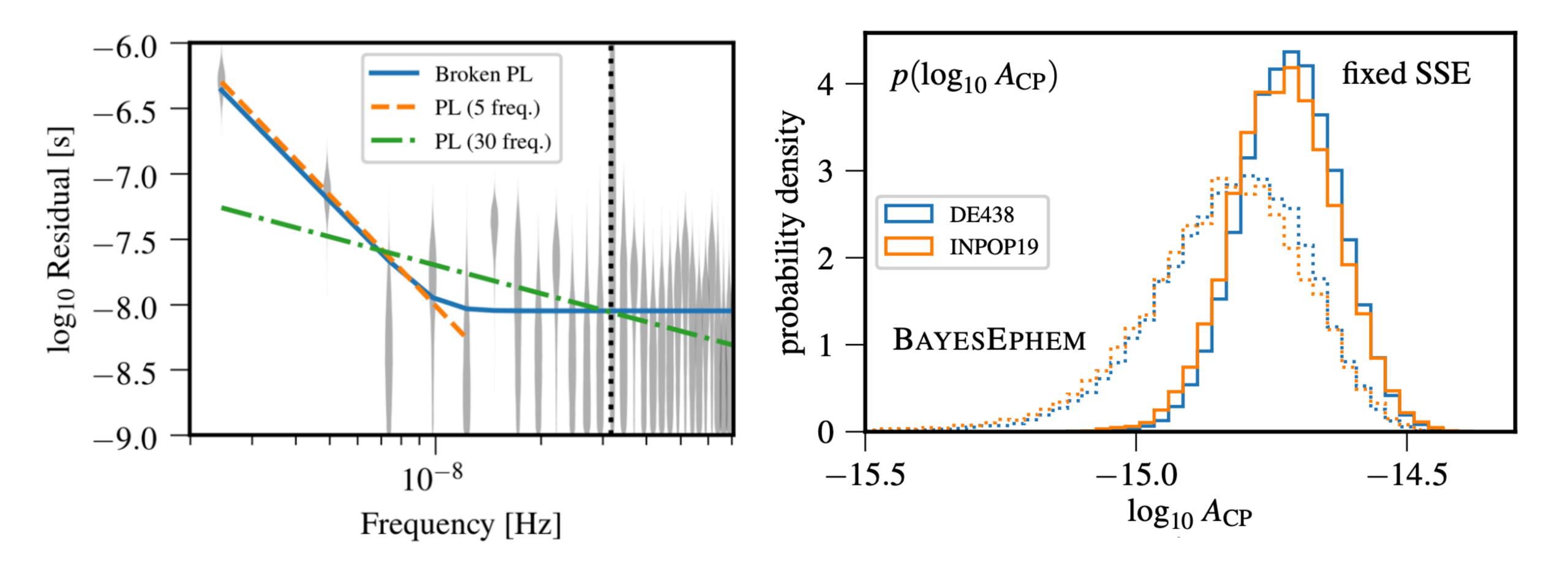
NANOGrav 12.5yr Dataset Search (arXiv:2009.04496), corresponding author: Joe Simon (JPL / CU-Boulder)



## A Common-spectrum Process

NANOGrav 12.5yr Dataset Search (arXiv:2009.04496),

corresponding author: Joe Simon (JPL / CU-Boulder)

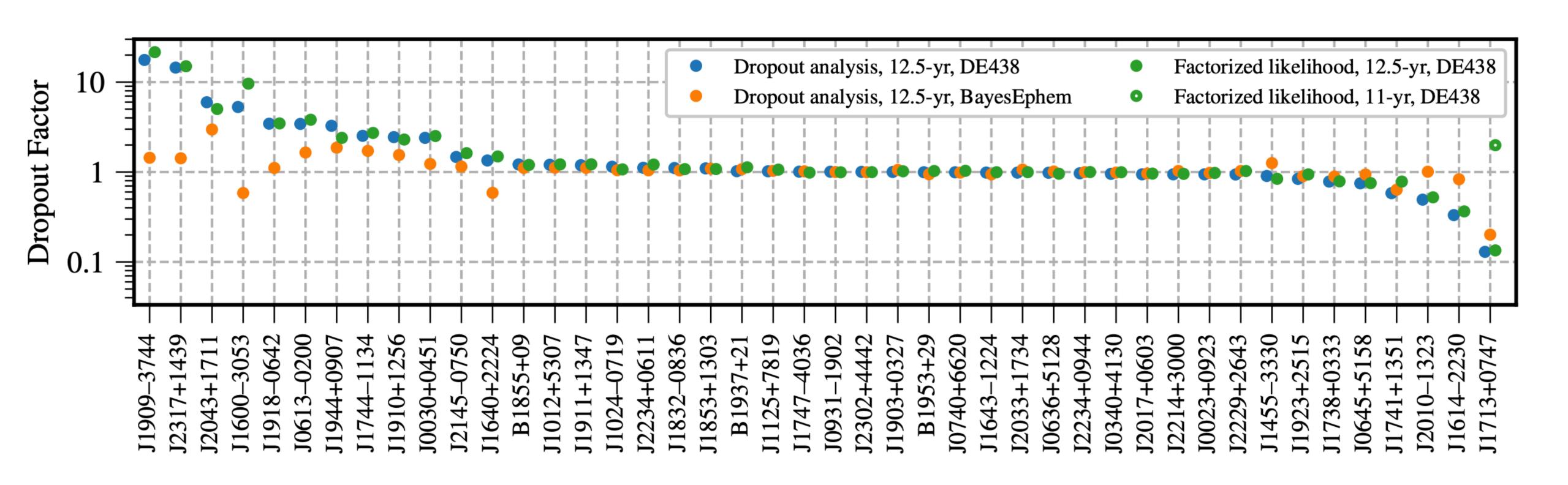


A steep-spectrum process in common across NANOGrav's 45-pulsar array with max baseline of 12.9 years

#### A Common-spectrum Process

NANOGrav 12.5yr Dataset Search (arXiv:2009.04496),

corresponding author: Joe Simon (JPL / CU-Boulder)



#### **Dropout factor = cross-validation probability**

i.e. how much does each pulsar support what is found by all other pulsars? S. Vigeland, S. Taylor, M. Vallisneri



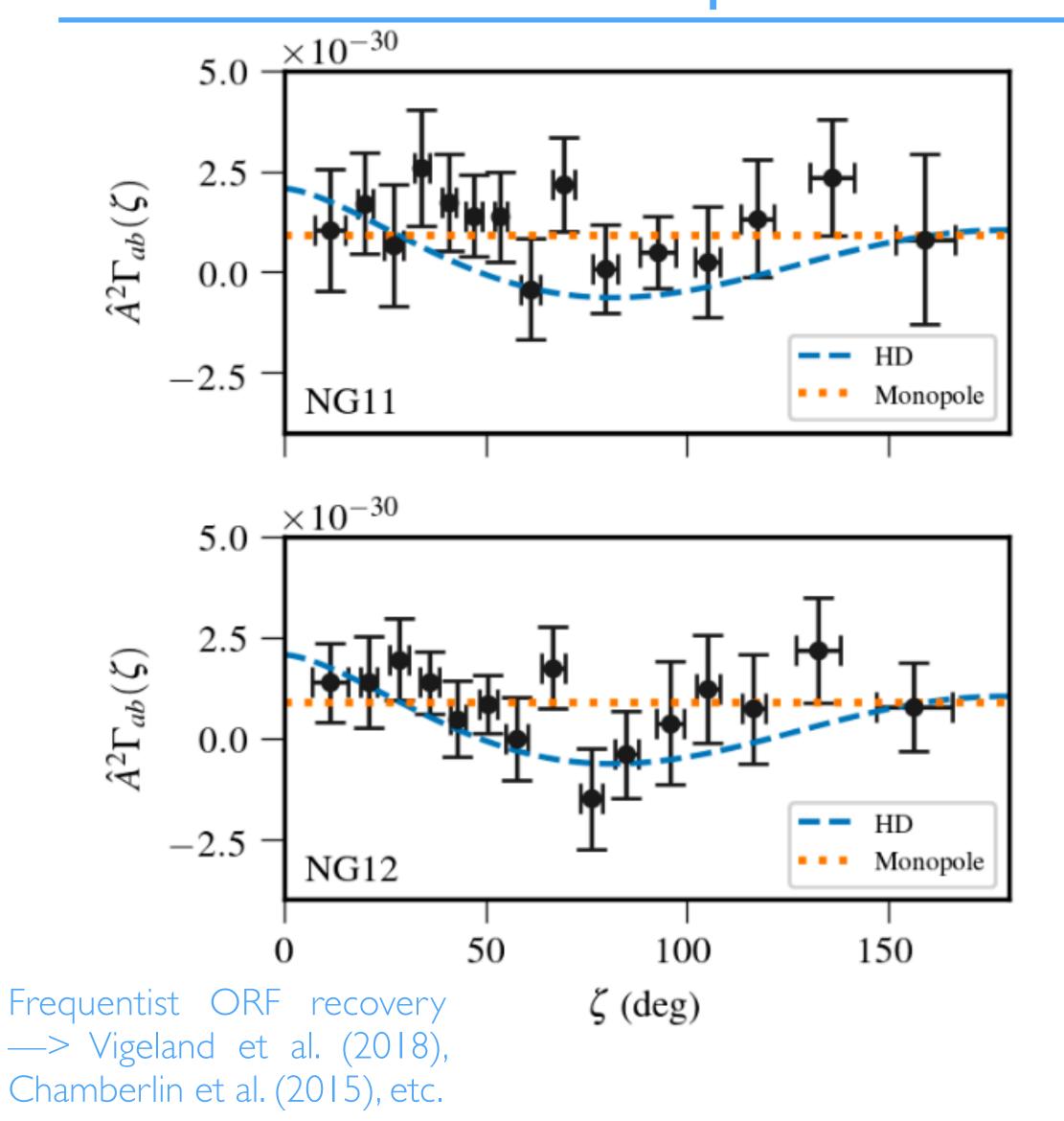


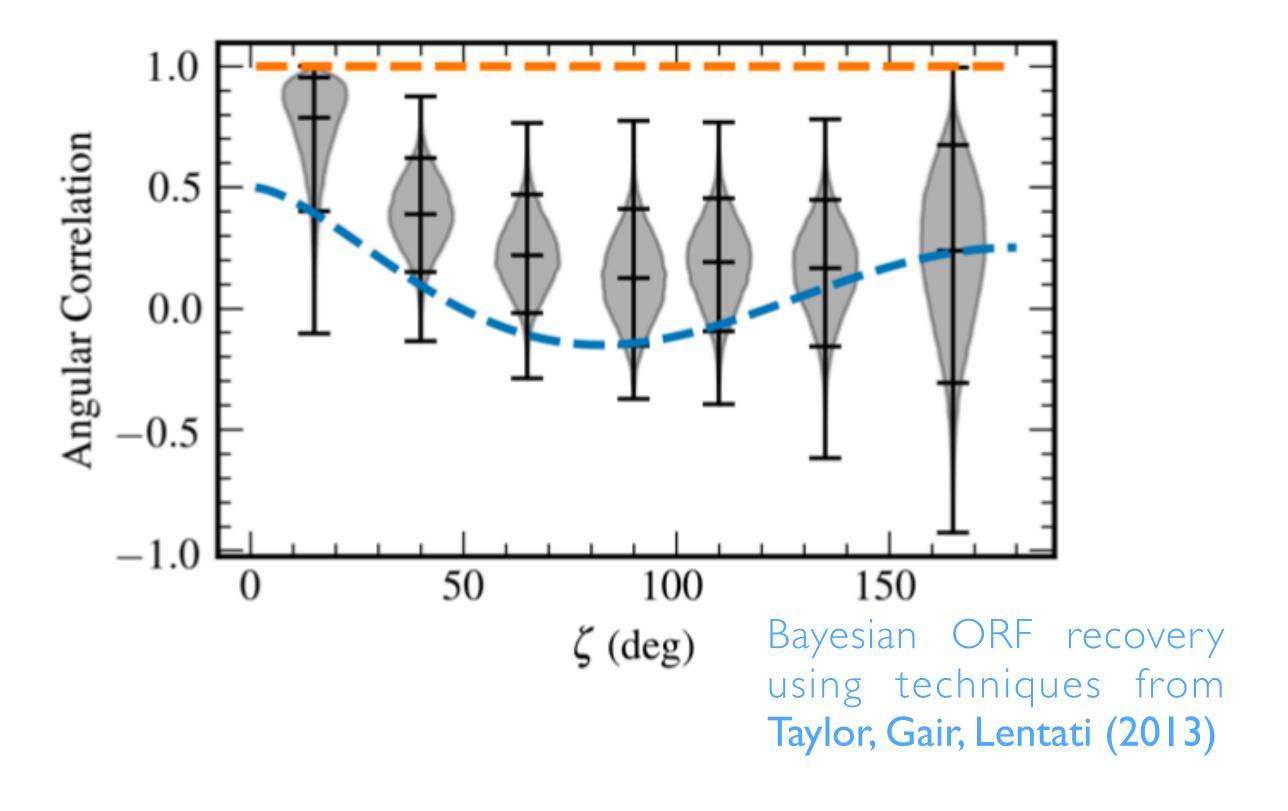


## A Common-spectrum Process

NANOGrav 12.5yr Dataset Search (arXiv:2009.04496),

corresponding author: Joe Simon (JPL / CU-Boulder)





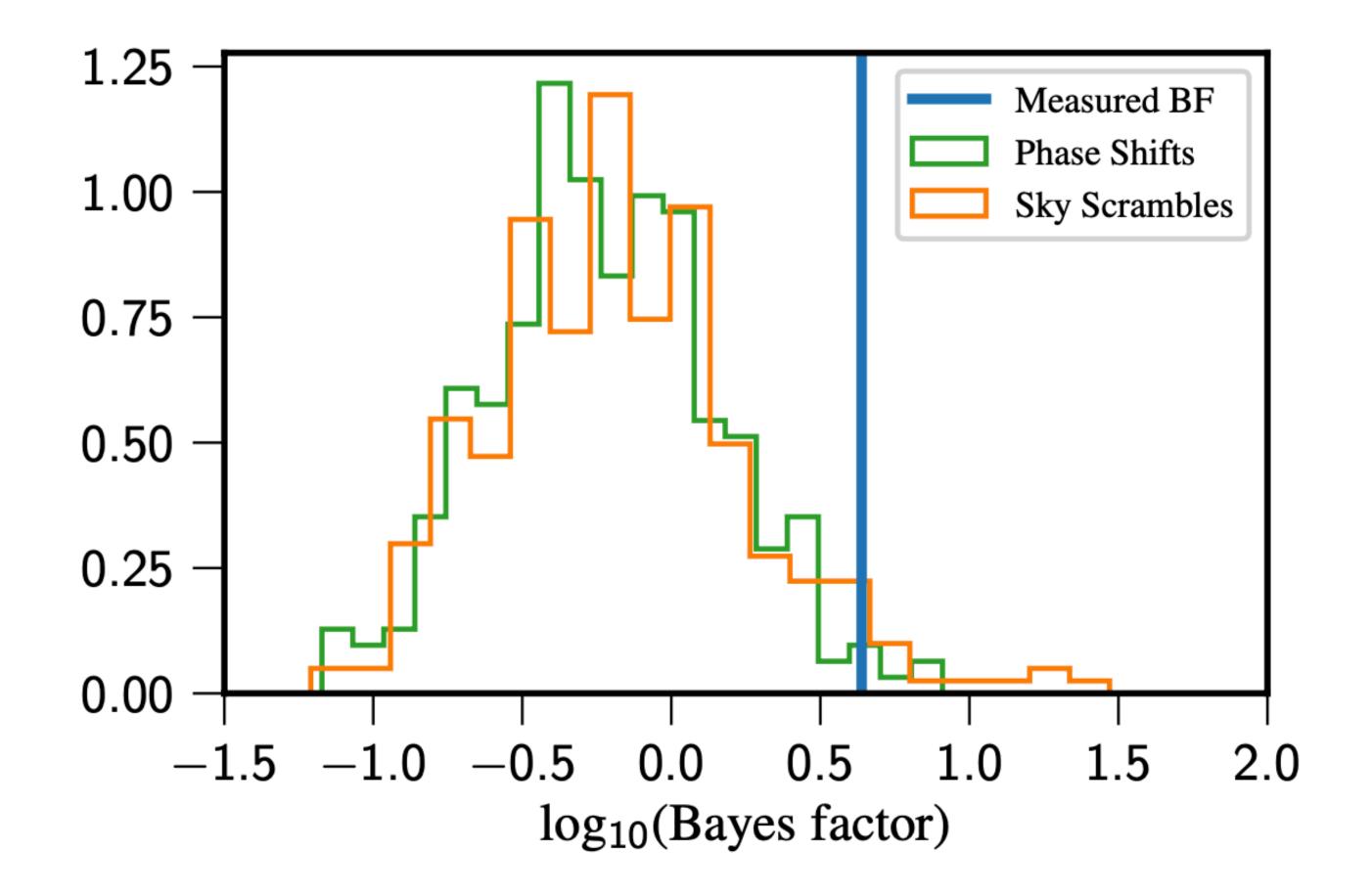
- Inter-pulsar correlations remain insignificant.
- Odds ratios for Hellings & Downs correlations
   ~2-4 depending on ephemeris modeling.





NANOGrav 12.5yr Dataset Search (arXiv:2009.04496),

corresponding author: Joe Simon (JPL / CU-Boulder)



- Assess the significance of spatial correlations by constructing null distribution.
- LIGO-Virgo-KAGRA use time slides... we use phase shifts (Taylor et al. 2017) and sky scrambles (Cornish & **Sampson 2016; Taylor et al. 2017).**
- p~5-10%

$$C_{gwb} = F \varphi_{gwb} F^T$$
Phase Shifting Sky Scrambles



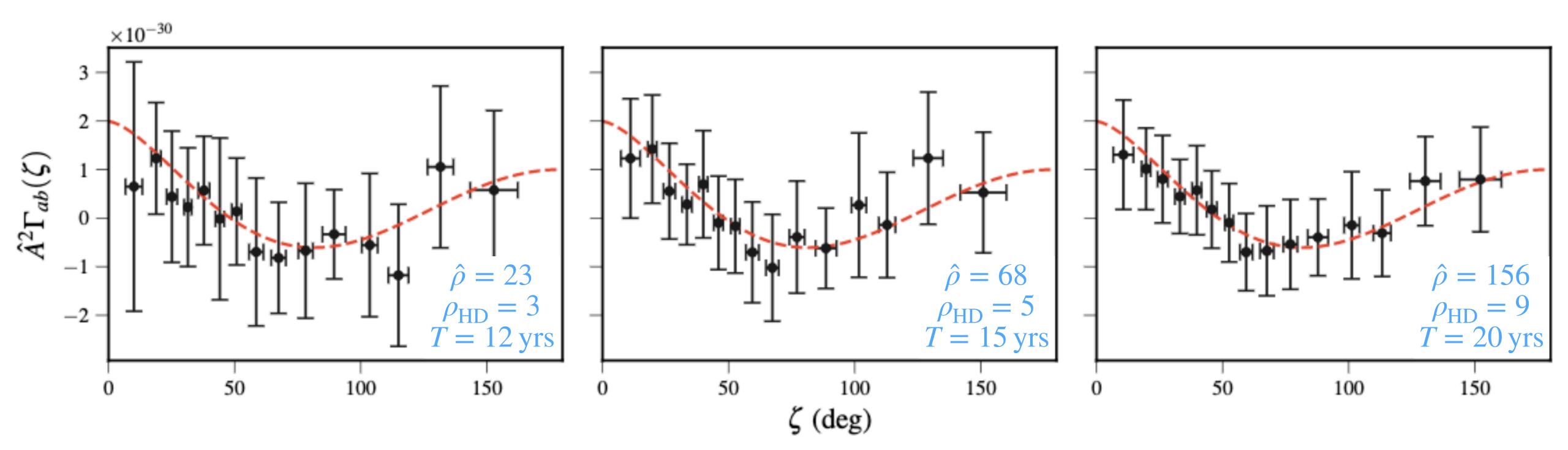
# The Road To & Beyond Detection

...Or "what to expect when you're expecting to detect a signal".



Dr. Nihan Pol

Simulate up to 20 years of PTA data, forecasting from the 45 pulsars in the NG 12.5yr data



$$\hat{\rho}=$$
 total S/N (from full log-likelihood ratio)  $\rho_{\rm HD}=$  cross-correlation S/N

Full team: Nihan Pol, Stephen Taylor, Luke Kelley, Joe Simon, Sarah Vigeland, Siyuan Chen

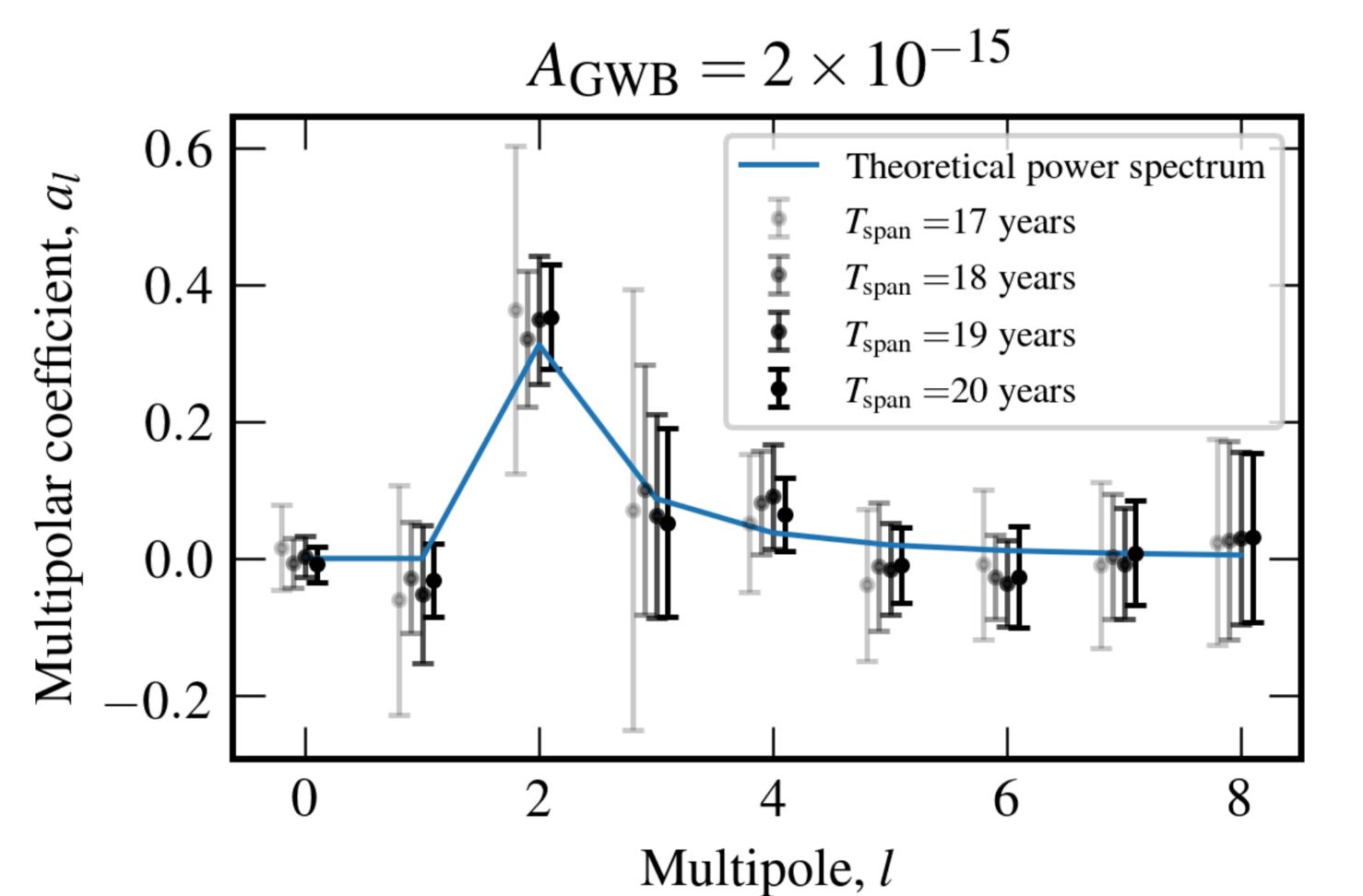


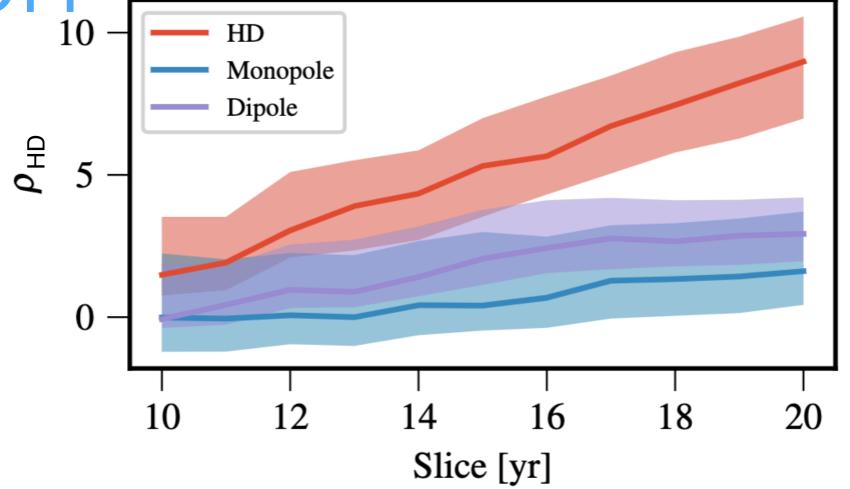


The Road To & Beyond Detection

...Or "what to expect when you're expecting to detect a signal".

Probe the multipolar structure of the inter-pulsar correlations

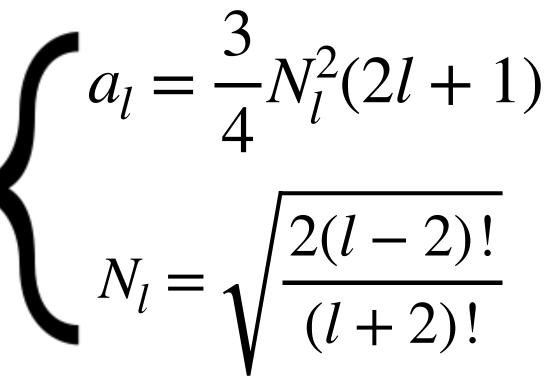




$$\Gamma_{ab} = \sum_{l=0}^{\infty} a_l P_l(\cos \theta_{ab})$$

Isotropic GWB:

Gair, Romano,
Taylor, Mingarelli (2014)







# The Road To & Beyond Detection

"Astrophysics Milestones For Pulsar Timing Array Gravitational Wave Detection", Pol, Taylor et al., arXiv:2010.11950

...Or "what to expect when you're expecting to detect a signal".

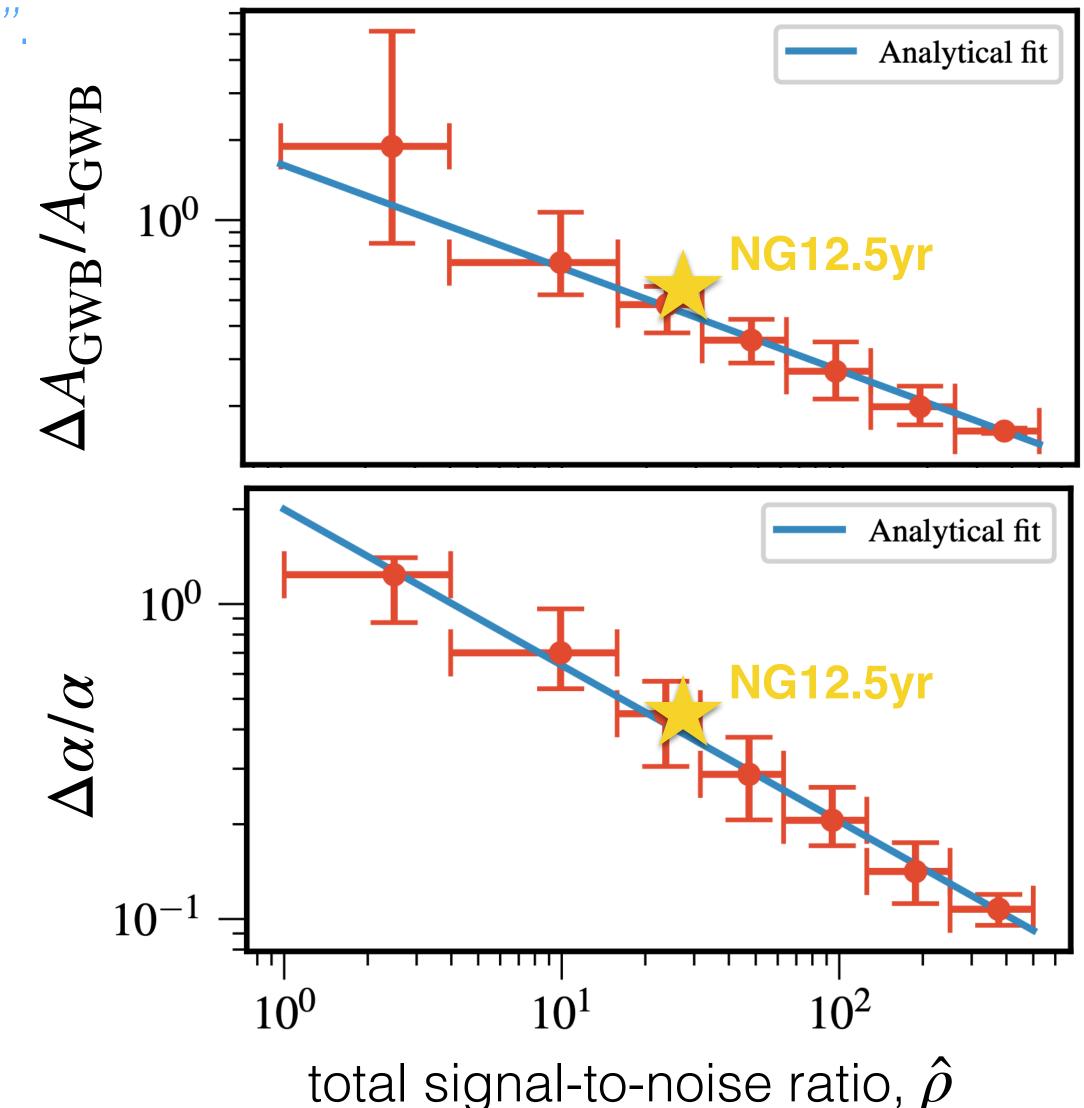
$$h_c(f) = A_{\text{GWB}} \left(\frac{f}{1 \text{ yr}^{-1}}\right)^{\alpha}$$

parameter uncertainty scaling laws

$$\Delta A_{\text{GWB}}/A_{\text{GWB}} = 44 \times \left(\frac{\hat{\rho}}{25}\right)^{-2/5} \%$$

$$\Delta \alpha/\alpha = 40 \times \left(\frac{\hat{\rho}}{25}\right)^{-1/2} \%$$

Can relate  $\hat{\rho}$  to  $\rho_{\mathrm{HD}}$  and factors like T,  $\sigma_{\mathrm{RMS}}$ ,  $N_{\mathrm{pulsar}}$ , etc.



# Summary

- · Pulsar Timing Arrays are sensitive to nanohertz gravitational waves.
- We use rank-reduced time-domain modeling of stochastic processes across dozens of pulsars and over decades of observations.
- If the NANOGrav result hints at a GWB, then **detection and characterization could be within a few years** (expedited by fusing datasets together in the IPTA).
- The road beyond detection will inform demographics and final-parsec binary dynamical interactions of supermassive binary black holes.



