

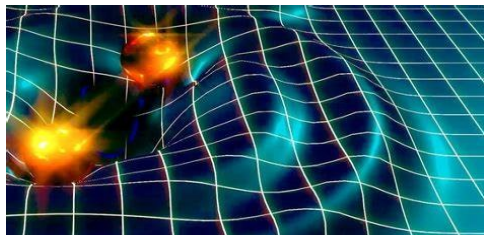
# Bootstrap for multivariate time series and gravitational wave detection

S.N. Lahiri

Department of Mathematics & Statistics  
Washington University in St. Louis

`s.lahiri@wustl.edu`

- As described by general relativity, GW are freely propagating wave solutions to Einstein's equation, or “ripples” in the space-time metric.



- GW are expected to be generated by nearly any configuration of accelerating mass.
- However, due to the weakness of gravity, large masses/high accelerations (e.g., binary systems of neutron stars) are required to radiate significant GW.

- GWs can be indirectly inferred using precise measurements of timings of radio pulses from spinning, magnetized neutron stars (pulsars).
- The pulse times of arrival can be analyzed via models incorporating the GW component.
- One operational model for the observed pulsar time of arrivals (TOAs) can be written as

$$\tau = \tau^{\text{TM}} + \tau^{\text{DM}} + \tau^{\text{GW}} + \tau^{\text{other}} \quad (0.1)$$

where

- $\tau^{\text{TM}}$ : Physical model for TOAs taking into account spin period, proper motion, binary orbital dynamics, etc.
- $\tau^{\text{DM}}$ : Model for time-varying dispersion measure variations.
- $\tau^{\text{GW}}$ : Model for any GWs. This includes stochastic sources that have a unique correlation pattern across multiple pulsars, etc!

- Through some clever manipulation & approximations (cf. Demorest et al. (2007, 2012)), this leads to the following model for the *pre-fit* residuals for a single pulsar:

$$\mathbf{Y} = \mathbf{A}\beta + \epsilon.$$

- Parameters  $\beta$  are estimated by WLS/GLS, giving

$$\hat{\beta} = (\mathbf{A}'\mathbf{W}\mathbf{A})^{-1}\mathbf{A}'\mathbf{W}\mathbf{Y}.$$

where  $\mathbf{W}$  is (often) a diagonal matrix with inverse variances at each epoch (but can be more general).

- The *post-fit* residuals are given by  $\mathbf{R}\mathbf{Y}$ , where  $\mathbf{R}$  is the projection operator:

$$\mathbf{R} = \mathbf{I} - \mathbf{A}(\mathbf{A}'\mathbf{W}\mathbf{A})^{-1}\mathbf{A}'\mathbf{W}.$$

- It is easy to check that  $\mathbf{R}$  is idempotent (and singular).
- Further,

$$\mathbf{R}\mathbf{Y} = \mathbf{R}\mathbf{A}\beta + \mathbf{R}\epsilon = \mathbf{0} + \mathbf{R}\epsilon.$$

- Thus, we can use the post-fit residuals to investigate the covariance structure of  $\epsilon$  that contains information about the GWB !!

- We write

$$\epsilon = \epsilon^{\text{gW}} + \epsilon^{\text{other}},$$

where  $\epsilon^{\text{gW}}$  denotes the part of the noise due to GWB.

- Under an isotropic power law spectrum *assumption*, the **GWB covariances** are of the form:

$$((\mathbf{C}^{\text{gW}}))_{ij} = A_1^2 C^{\text{gW}}(t_i - t_j)$$

where

- $C^{\text{gW}}(\cdot)$  is the covariance function corresponding to the spectrum and
- $A_1$  is the unknown GW spectrum amplitude at the reference frequency  $f_0 = 1\text{yr}^{-1}$  - **the parameter of interest!!!**

- The presence of GW component make the residual series of different pulsars correlated!!
- For a pair of pulsars  $(a, b)$ , the noise variables  $\epsilon_a$  and  $\epsilon_b$  can be written as

$$\epsilon_a = \epsilon_a^{\text{gw}} + \epsilon_a^{\text{other}}, \quad \epsilon_b = \epsilon_b^{\text{gw}} + \epsilon_b^{\text{other}}$$

where  $\epsilon_a^{\text{gw}}$  and  $\epsilon_b^{\text{gw}}$  are correlated, but  $\epsilon_a^{\text{other}}$  and  $\epsilon_b^{\text{other}}$  are NOT!

# The GW component in residuals

- **The covariance between the noise variables  $\epsilon_a$  and  $\epsilon_b$  for a pair of pulsars  $(a, b)$  is given by**

$$((\mathbf{C}_{a,b}))_{ij} = A_1^2 C^{\text{GW}}(t_i - t_j) \zeta(\theta_{ab})$$

where

- $\theta_{ab}$  = the angular separation between pulsars  $a$  and  $b$  and
- $\zeta(\cdot)$  is the *Hellings-Downs function* !
- Thus, the cross-covariance matrix between  $\epsilon_a$  and  $\epsilon_b$  is determined by the GW power spectrum.
- Further, in  $\mathbf{C}_{a,b}$ , **the ONLY unknown parameter is  $A_1^2$ .**

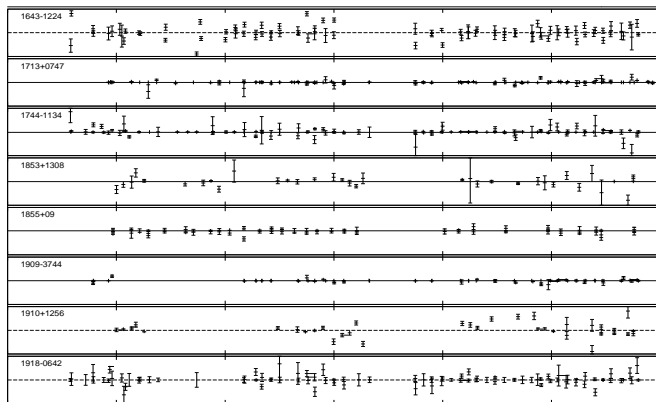


# The GW component in residuals

Q: How do we estimate  $A_1$ ?

# Cadence of TOAs/ residuals

- Here is a plot of the residuals for different pulsars showing their cadence (cf. Demorest et al. (2012)):



Note the following features:

- The time points are **irregularly** spaced!
- The **number of observations can be/are different** for any two distinct pulsars!
- The coverage is **NOT uniform** and there are **gaps appearing during 2007** in most of the series!
- The **densities and spans** of different series can be very **different!**
- There is **heteroskedsticity** (cf. size of the bars) !!

# Optimal Statistic :

- Demorest et al. (2012) defined the following cross-correlation statistic (cf. eqn (9), p.13):

$$\rho_{ab} = \frac{\sum_{ijkl} r_i^{(a)} (C^{tot(a)})_{ij}^- C_{jk}(a, b) (C^{tot(b)})_{kl}^- r_l^{(b)}}{\sum_{ijkl} (C^{tot(a)})_{ij}^- C_{jk}(a, b) (C^{tot(b)})_{kl}^- C_{il}(a, b)}$$

where

- $i, j \in \{1, \dots, N_a\}$  and  $k, l \in \{1, \dots, N_b\}$ , with  $N_a$  denoting the number of TOAs for pulsar  $a$ , etc.
- $r_i^{(a)}$  and  $r_l^{(b)}$  are the post-fit timing residuals for pulsars  $a$  and  $b$ , respectively.
- $\mathbf{C}^{tot(a)}$  is the (estimated) covariance matrix of the post-fit residuals for pulsar  $a$ , and  $(\mathbf{C}^{tot(a)})^-$  is its generalized inverse!
- $C_{il}(a, b)$  is the  $(i, l)$  element of  $R_a[\mathbf{C}_{a,b}]R_b'$ .

- It can be shown that for all  $a, b$ ,

$$E\rho_{ab} = A_1^2 \zeta(\theta_{ab}).$$

- Given a set of pulsars  $\{1, \dots, m\}$ , we can set up the regression model:

$$\rho_{ab} = A_1^2 \zeta(\theta_{ab}) + \epsilon_{ab}$$

for pairs  $(a, b) \in \Gamma$ , where  $\Gamma = \{(a, b) : 1 \leq a < b \leq m\}$ .

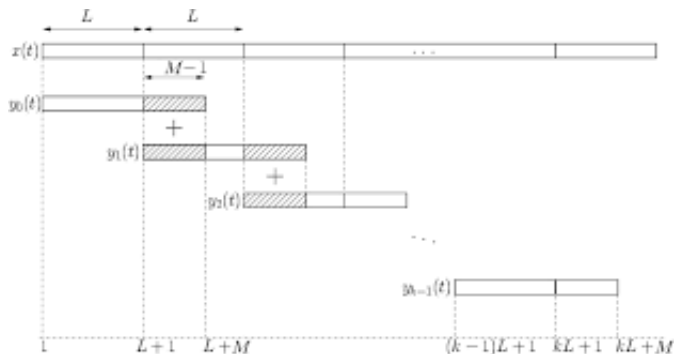
- This leads to an estimator of  $A_1^2$  of the form:

$$\hat{A}_1^2 = \frac{\sum_{(a,b)} \zeta(\theta_{ab}) \rho_{ab}}{\sum_{(a,b)} \zeta(\theta_{ab})^2}.$$

**Q:** How do we approximate the distribution of  $\hat{A}_1^2$ ?

- We can use the Bootstrap to approximate the distribution of  $\hat{A}_1^2$ .
- But what form of Bootstrap is appropriate?
  - Sampling with replacement / IID Bootstrap ?
  - Block Bootstrap ?

# Construction of the Blocks for time series



- $M = 1$  gives the maximum overlapping version
- $M > 1$  can be used to reduce computational burden



# Issues with the Block Bootstrap

- While this will work with a single time series, it may not be effective under the present scenario:
  - the number of pulsars  $\approx 37+$
  - and the sample size is only around 500+.
- Thus, - **curse of dimensionality** will kick in!!
- Time-domain Block Bootstrap does not handle Red Noise very well (cf. Lahiri (1993)).
- There are also issues with irregularly spaced time-points! (cf. Lahiri and Zhu (2006)).

# A new Bootstrap method

- Recall that

$$\rho_{ab} = \frac{\sum_{ijkl} r_i^{(a)} (C^{\text{tot}(a)})_{ij}^- C_{jk}(a, b) (C^{\text{tot}(b)})_{kl}^- r_l^{(b)}}{\sum_{ijkl} (C^{\text{tot}(a)})_{ij}^- C_{jk}(a, b) (C^{\text{tot}(b)})_{kl}^- \tilde{C}_{il}(a, b)}.$$

- Define

$$\mathbf{Z}^{(a)} \equiv (Z_1^{(a)}, \dots, Z_{N_a}^{(a)})' = (\mathbf{C}^{\text{tot}(a)})^{-1/2} \mathbf{r}^{(a)},$$

the set of pre-whitened residuals for pulsar  $a$ .

- Note that  $\rho_{ab}$  can be written as

$$\rho_{ab} = \sum_{i=1}^{N_a} \sum_{l=1}^{N_b} w_{ab}(i, l) Z_i^{(a)} Z_l^{(b)}$$

for some weights  $w_{ab}(i, l)$ .

# A new Bootstrap method

- Thus, to define the Bootstrap version of

$$\hat{A}_1^2 = \frac{\sum_{(a,b)} \zeta(\theta_{ab}) \rho_{ab}}{\sum_{(a,b)} \zeta(\theta_{ab})^2} = \sum_{(a,b)} \sum_{i=1}^{N_a} \sum_{l=1}^{N_b} \tilde{w}_{ab}(i, l) Z_i^{(a)} Z_l^{(b)},$$

it is enough to be able to generate Bootstrap versions of  $\mathbf{Z}^{(a)}$  for all  $a$ .

- Note that for each  $a$ , the variables  $Z_1^{(a)}, \dots, Z_{N_a}^{(a)}$  are approximately iid.
- So, we can resample with replacement to generate the (pre-)Bootstrap sample

$$Z_1^{o(a)}, \dots, Z_{N_a}^{o(a)}.$$

# A new Bootstrap method

- Note that this direct resampling only captures the **marginal** behavior of each set of residuals.
- It is important to also capture their interactions (correlations)!
- Thus, the Bootstrap version of  $\{\mathbf{Z}^{(a)} : a = 1, \dots, m\}$  is defined by

$$\begin{pmatrix} \mathbf{Z}^{*(a)} \\ \dots \\ \mathbf{Z}^{*(m)} \end{pmatrix} = \Sigma^{1/2} \begin{pmatrix} \mathbf{Z}^{\circ(a)} \\ \dots \\ \mathbf{Z}^{\circ(m)} \end{pmatrix}$$

where  $\Sigma$  is an  $N \times N$  matrix (with  $N = N_1 + \dots + N_m$ ) consisting of  $m \times m$  block matrices with  $(a, b)$ th block :

$$[\mathbf{C}^{tot(a)}]^{-1/2} \mathbf{C}_{a,b} [\mathbf{C}^{tot(b)}]^{-1/2}, \quad a \neq b.$$

# A new Bootstrap method

- Now define

$$[A_1^*]^2 = \sum_{(a,b)} \sum_{i=1}^{N_a} \sum_{l=1}^{N_b} \tilde{w}_{ab}(i, l) Z_i^{*(a)} Z_i^{*(b)}.$$

- We can use the Bootstrap distribution of  $[A_1^*]^2$  to approximate the distribution of  $\hat{A}_1^2$ .
  - We can use the Monte-Carlo based Bootstrap quantiles of  $[A_1^*]^2$  to construct CIs for  $A_1^2$ .
  - This, in turn, can be used for testing  $H_0 : A_1 = 0$ , etc.

# The END!!!

**Thank you !!**