Probing the Large-scale Structure of the Universe Through Gravitational Wave Observations

Xiaoyun Shao1, Zhoujian Cao1, Xilong Fan2, and Shichao Wu1,3,4
1 Department of Astronomy, Beijing Normal University, Beijing 100875, China
2 School of Physics and Technology, Wuhan University, Wuhan 430072, China; xilong.fan@whu.edu.cn
3 Max-Planck-Institut für Gravitationsphysik (Albert-Einstein-Institut), D-30167 Hannover, Germany
4 Leibniz Universität Hannover, D-30167 Hannover, Germany

Received 2021 July 23; revised 2021 September 30; accepted 2021 October 14; published 2022 January 21

Abstract

The improvements in the sensitivity of the gravitational wave (GW) network enable the detection of several large redshift GW sources by third-generation GW detectors. These advancements provide an independent method to probe the large-scale structure of the universe by using the clustering of the binary black holes (BBHs). The black hole catalogs are complementary to the galaxy catalogs because of large redshifts of GW events, which may imply that BBHs are a better choice than galaxies to probe the large-scale structure of the universe and cosmic evolution over a large redshift range. To probe the large-scale structure, we used the sky position of the BBHs observed by third-generation GW detectors to calculate the angular correlation function and the bias factor of the population of BBHs. This method is also statistically significant as 5000 BBHs are simulated. Moreover, for the third-generation GW detectors, we found that the bias factor can be recovered to within 33% with an observational time of ten years. This method only depends on the GW source-location posteriors; hence, it can be an independent method to reveal the formation mechanisms and origin of the BBH mergers compared to the electromagnetic method.

Key words: gravitational waves – (cosmology:) large-scale structure of universe – black hole physics

1. Introduction

Gravitational waves (GWs) were successfully detected by the Laser Interferometer Gravitational Wave Observatory (LIGO) on 2015 September 15 (Abbott et al. 2016c). There have been several black hole binaries detected by the Advanced LIGO and Virgo interferometers (Acernese et al. 2014; Aasi et al. 2015; Abbott et al. 2016a, 2019a, 2019b, 2019c, 2020a, 2020b, 2020c). The First GW Transient Catalog (GWTC-1) from the first two observing runs reported 11 GW events from compact binary mergers (Abbott et al. 2019b), which was published by the LIGO-Virgo Collaboration. The Second GW Transient Catalog (GWTC-2) includes a total of 44 confident binary black hole (BBH) events (Abbott et al. 2021a, 2021b). With the increasing number of detected events, the observation of GWs has heralded a new era.

Based on the LIGO-Virgo detectors, most GW signals originate from the merger of compact objects (Acernese et al. 2014). In coming years, KAGRA (Aso et al. 2013) and LIGO-India (Iyer et al. 2011) could join a network of GW observatories. The proposals of the third-generation detectors (3G) greatly improve the sensitivity of a network of detectors. The proposed 3G detectors include LIGO Voyager (Adhikari et al. 2019, L. S. Collaboration et al. 2015), Einstein Telescope (ET) (Punturo et al. 2010, L. S. Collaboration et al. 2015), and Cosmic Explorer (CE) (Dwyer et al. 2015, L. S. Collaboration et al. 2015).

GWs can become a new way to probe cosmology in the future. The Hubble constant is measured following the multi-messenger detection of the binary neutron star merger GW170817 with a measured value $H_0 = 70.0^{+12.0}_{-8.0} \text{ km s}^{-1} \text{ Mpc}^{-1}$ (Abbott et al. 2017). By using the multi-messenger method, 50 detections of binary neutron star mergers by the Advanced LIGO and Virgo interferometers may lead to a precise measurement of the Hubble constant (Feeney et al. 2019; Chen et al. 2020) and resolve its inconsistent measurements (Schutz 1986; Abbott et al. 2017).

We can use the millions of BBHs observed by future GW detectors as tracers of large-scale structures over a large redshift range. The large-scale structure is studied by using the number density and luminosity distances observed by future GW detectors (Namikawa et al. 2016a; Namikawa 2021). The studies find that the distribution of BBHs mergers within GWTC-1 is isotropic and the anisotropy can be probed with an angular scale of degrees at the level of $<= 0.1\%$ within an observational time of a year (Payne et al. 2020). The tracer with BBHs is independent and complementary to the tracer with galaxy clustering. The black hole catalogs are complementary to the galaxy catalogs because of large redshifts of GW events ($z$ is up to 10) (L. S. Collaboration 2019), which may suggest that BBHs are a better choice than galaxies to probe the large-scale structure of the universe and cosmic evolution over a large redshift range. Beyond this aspect, we can characterize...
their clustering properties and trace dark matter distribution with cosmological bias by statistically studying the spatial distribution of GW events. The matter inhomogeneities in the universe can be traced by the spatial distribution of BBHs. Its results show that the clustering bias of BBHs observed by the second-generation GW detectors are greater than or equal to 1.5 with an observational time of three years (Namikawa et al. 2016b). GW bias, which provides information on the physical nature of the different mergers, can be measured by cross-correlation between the galaxy and GW surveys (Raccanelli et al. 2016; Banagiri et al. 2020; Calore et al. 2020). Therefore, we can determine whether black hole mergers trace the distribution of dark matter more closely than other traditional methods, which may suggest a primordial origin of BBHs, or that of stars harbored in luminous and massive galaxies, leading to a stellar origin of BBHs (Scelfo et al. 2018). Another approach to measuring GW bias based on source-location posteriors and a two-point correlation function is proposed (Vijaykumar et al. 2020).

Although statistics can be derived from merging events of double black holes for constructing the large-scale structure of the universe based on GW detection (Camera & Nishizawa 2013; Namikawa et al. 2016a, 2016b; Calore et al. 2020; Namikawa 2021), the position accuracy of a GW source measured by a ground-based detector is poor. The error in the angular direction is more than 10 square degrees and that of the distance is more than 10% (Ashton et al. 2020). Even if an ideal space detector network is used, the error in the angular direction could exceed 1 square arcsec, and the error of the distance could exceed 10% (Ruan et al. 2020; Hu et al. 2021; Zhang et al. 2021). Therefore, the feasibility to construct the large-scale structures of the universe by GW detection is questioned.

Recently, a research group has neglected the correlations between the errors in R.A., decl., and distance. Based on this assumption, the team concluded that the third-generation GW detectors can build large-scale structures of the universe as long as the error in the angular direction is less than 1 square degree and the error of distance is less than ∼90 h−1 Mpc (Vijaykumar et al. 2020). We rejected the above assumption to calculate more accurate observations. We assume that the position measurement errors in the three directions of space are correlated, and the value of the error depends on the distance of the GW source in our present study. We use the angular correlation function (ACF) to explore the effect of estimating the bias factor from the spatial distribution of the observed BBH population, which is useful in identifying the GW events’ origins.

This paper is organized as follows: we describe how to theoretically calculate the smeared ACF, where the relationship between the ACF and the smearing ACF is derived in Section 2. We detail the simulation of the BBH catalogs and estimate the bias factor by using the least square method in Section 3. We discuss the results in Section 4.

2. Theory

2.1. Angular Correlation Function (ACF)

The ACF ω(θ) is defined as the excess probability in comparison with a random Poisson distribution, which can be expressed as a function of angular separation θ on the sky. We can express the ACF through the two-point correlation function ξ(R) using the Limber approximation (Simon 2007)

\[ w(\theta) \approx \int_0^\infty dr_1 \int_0^\infty dr_2 p_1(r_1) p_2(r_2) \xi(R) \]

where

\[ R \equiv \sqrt{r_1^2 + r_2^2 - 2 r_1 r_2 \cos \theta} \]

and

\[ P_1 = P_2 = P(r) \begin{cases} 0 & 0 \leq r < r_1 \\ 1 & r_1 \leq r < r_2 \\ 0 & r \geq r_2 \end{cases} \]

where \( r_1 \) and \( r_2 \) are the distances of two GW events from the observer, and \( R \) is the distance separation between the two events.

The observed BBH population can be used to reveal traces of dark matter because dark matter is more abundantly found than light-emitting matter (Aghanim et al. 2019). The ability to trace dark matter can be quantitatively expressed as (Vijaykumar et al. 2020)

\[ \xi_{BBH}(r) = b_{BBH}^2 \xi_{DM}(r). \]

With the absence of measurement errors, we can then use the ACF of dark matter and the bias factor of BBH to represent the ACF of BBHs, which is derived by using Equations (1) and (4), and is given by

\[ \omega_{BBH}(\theta) = b_{BBH}^2 \omega_{DM}(\theta). \]

2.2. Estimating the ACF

The Landy–Szalay (LS) estimator can estimate the ACF from the data of the population of points according to the definition of \( \omega(\theta) \), that is, the excess probability of seeking out two points with an interval of θ as compared to a random Poisson distribution, and is given by

\[ \omega(\theta) = \frac{DD(\theta) - 2DR(\theta) + RR(\theta)}{RR(\theta)}. \]

where \( DD(\theta) \), \( DR(\theta) \), and \( RR(\theta) \) are the number of data–data, data-random and random–random pairs normalized by the number of random–random pairs in each angular bin, respectively, Landy & Szalay (1993).
Several BBH mergers could be observed by using ET and CE detectors; thus, the clustering of BBHs could be explored by estimating \( \omega(\theta) \) with the BBHs spatial distribution. However, the precision in the GW source location could be poor as compared to the galaxies location (which can be described as a point in the survey volume). Therefore, the observed ACF of BBHs must be different from the true ACF because of the large measurement error of the GW source location, implying that the poor source localization distributes weights from the points of the actual location to a smeared field around those points. The distribution of the GW source localization uncertainties from the population affects the smeared ACF. By convolving ACF with the GW source-location posteriors, the smeared ACF can be computed (Figure 1).

This study assumes that the two-dimensional source-location posterior distribution of the GW event satisfies the Gaussian distribution with its true position as the mean value, and the error matrix is proportional to the square of the distance from the GW source. Future GW detectors with high sensitivity can perform reasonably well because the measurement error is inversely proportional to the square of the signal-to-noise ratio (S/N), and the S/N is inversely proportional to the distance from the GW source (Finn & Chernoff 1993). A normalized total posterior distribution in R.A. and decl. \( P(p) = \frac{1}{\sqrt{2\pi\sigma_1\sigma_2}} \exp\left[-\frac{1}{2}\left(p - m_l - \Delta m_l\right)^T\right] \times C_l^{-1}\left(p - m_l - \Delta m_l\right) \]

where \( p = (\alpha, \beta) \), \( \alpha \) is R.A., \( \beta \) is decl., and \( m_l \) is the true location in two-dimensions of the \( l \)th BBHs. The covariance matrix of the Gaussian distribution of \( l \)th GW source location is denoted as \( C_l \), while \( N = 1/\sqrt{(2\pi)^2|C_l|} \) is a normalized factor related to distance. The detector noise introduces the scatter \( \Delta m_l \) so that the individual GW event posterior is not centered on the true location. \( \Delta m_l \) satisfies the Gaussian distribution of the zero-mean and the error matrix \( C_l \).

The smeared ACF can then be calculated via the LS estimator after sampling from the Gaussian distribution. After repeating these steps many times (see Section 3.2), we can obtain the average value and error bars of smeared \( \omega_{\text{BBH}} \). The software package Corrfunc (Sinha & Garrison 2020) is used to estimate the smeared black hole ACF. It is contrasted with the smeared black hole ACF deduced by the theoretical dark matter correlation function and bias factor (see Section 2.4), so that the bias factor \( b_{\text{BBH}} \) can be recovered.

### 2.3. The Distance Effect

For more accurate observational GW source-location posteriors detected by future GW detectors, we assume that the error matrix of the GW source-location posterior is proportional to the square of the distance from the GW source as the measurement error is inversely proportional to the square of the S/N, and the S/N is inversely proportional to the distance from the GW source. The correlations between the errors in R.A. \( \alpha \) and decl. \( \beta \) are also considered.

In this study, GW source-location posteriors are assumed to be a Gaussian distribution with a covariance matrix \( C_l \), which is given by

\[
P_l(p - m_l, \Delta m_l, d_l) = \frac{1}{\sqrt{2\pi|C_l|}} \exp\left[-\frac{1}{2}(p - m_l - \Delta m_l)^T \right] \times C_l^{-1}(p - m_l - \Delta m_l).
\]

The covariance matrix \( C_l \) depends on the S/N and source sky location \((\alpha, \beta)\). Next, we deduce the relationship between the covariance matrix of the GW source-location posterior and the distance from the GW source. For simplicity, we only deduce it in the situation when the covariance matrix is diagonal; however, our conclusion remains valid when the covariance matrix is non-diagonal. The diagonal matrix is written as,

\[
C = \begin{pmatrix}
\sigma_1^2 & 0 \\
0 & \sigma_2^2
\end{pmatrix}
\]

where \( \sigma_1 \) and \( \sigma_2 \) are, respectively, 1σ error of the posterior distributions of sky location in R.A. and decl. The errors in the location \( \Delta \Omega \) can be written as,

\[
\Delta \Omega \propto \sqrt{\sigma_1^2 \sigma_2^2}
\]
\[ \Delta \Omega \propto \frac{1}{\rho^2}. \]  

(11)

Here we ignore the sky location dependence and relate the S/N to the distance:

\[ \rho^2 \propto \frac{1}{d^2}, \]  

(12)

where \( \rho \) is the S/N, and \( d \) is the distance from the GW source. Therefore we can derive the following:

\[ \sigma_1^2 \propto d^2 \]  

(13)

\[ \sigma_2^2 \propto d^2, \]  

(14)

Then we can verify that the covariance matrix of the GW source-location posterior is proportional to the square of the distance from the GW source, which is written as

\[ C \propto d^2. \]  

(15)

Thus, the covariance matrix for the individual GW source is given by:

\[ C_l = a \ast \left( \frac{d_l}{d_0} \right)^2 \ast C_0, \]  

(16)

where \( C_0 \) is a reference covariance matrix for the GW source at the distance \( d_0 \), and \( a \) is scaled by the sensitivity improvements in the detectors.

### 2.4. The Smeared Theoretical Black Hole ACF

We calculate the true ACF of BBH mergers before deducing the smeared ACF. The total source-location posterior in the two-dimensions of BBH mergers with no measurement error is denoted as

\[ P_0(m) = \frac{1}{N} \sum_l \delta^{(2)}(m - m_l), \]  

(17)

where \( \delta^{(2)} \) and \( m_l \) respectively, denote the two-dimensional Dirac delta function and the sky location \((\alpha_l, \beta_l)\) of the \( l \)th BBH. \( S \) is the area in two-dimensions including all possible values of R.A. and decl. The number of BBHs is denoted as \( N \). The density in excess \( \delta_0 \) is written as

\[ \delta_0(m) = P_0(m) / P_0 - 1 = S P_0(m) - 1. \]  

(18)

The true ACF is written as

\[ \omega_0(m, n) = \langle \delta_0(m) \delta_0(n) \rangle = S^2 (P_0(m) P_0(n)) - 1, \]  

(19)

where \( m \) and \( n \) are two independent points. \( P_0 = \frac{1}{3} \) is the area-averaged probability and \( \langle \cdot \rangle \) denotes ensemble averages. By using Equation (17), \( \langle P_0(m) P_0(n) \rangle \) can be written as

\[ \langle P_0(m) P_0(n) \rangle = \frac{1}{N^2} \left( \sum_{m, n} \delta^{(2)}(m - m_l) \delta^{(2)}(n - n_l) \right) . \]  

(20)

We calculate the smeared ACF of BBHs mergers with location measurement errors. The individual GW source-location posterior is considered as a Gaussian distribution, which is given by

\[ P_l(m - m_l, \Delta m_l, d_l) = \frac{1}{\sqrt{(2\pi)^2 |C_l|}} \]  

\[ \times \exp \left[ -\frac{1}{2}(m - m_l - \Delta m_l)^T \right. \]  

\[ \times C_l^{-1}(m - m_l - \Delta m_l), \]  

(21)

where \( m_l \) and \( C_l \), respectively, denote the true sky location and the covariance matrix for the \( l \)th GW source-location posterior. We assume that the posterior is proportional to the square of the distance from the GW source \( d_l \), and the scatter \( \Delta m_l \) induced by the detector noise meets the Gaussian distribution of mean value zero and covariance matrix \( C_l \). We first marginalize \( P_l(m - m_l, d_l, \Delta m_l) \) over \( \Delta m_l \):

\[ P_l(m - m_l, d_l) = \int dV_{\Delta m} P_l(\Delta m_l) \]  

\[ \times P_l(m - m_l, \Delta m_l, d_l). \]  

(22)

Then we marginalize \( P_l(m - m_l, d_l) \) over \( d_l \):

\[ P_l(m - m_l) = \int dV_{d} P_l(d_l) P_l(m - m_l, d_l) \]  

\[ = \text{Ad}^2, \quad r_1 \leq d_l < r_2, \]  

(23)

where \( r_1 \) and \( r_2 \) are radii of a survey spherical shell area and \( A \) is the normalization factor. The averages in Equations (22) and (23) can be calculated for the posterior (as opposed to the final ACF) only when the noise-induced shifts \( \Delta m_l \) are uncorrelated with the two-dimensional locations \( m_l, m_c \) can also be denoted as \( (\alpha_l, \beta_l) \). \( P_l(p - m_l) \) is written as

\[ P_l(p - m_l) = \int dS_m P_l(p - m_l) \delta^{(2)}(m - m_l). \]  

(25)

The total posterior of the sky location of the GW events is written as

\[ P(p) = \frac{1}{N} \sum_l P_l(p - m_l). \]  

(26)

In this probability field, the ACF is given by

\[ \langle P(p)P(q) \rangle = \frac{1}{N^2} \left( \sum_{l,n} P_l(p - m_l) \right. \]  

\[ \times \left. P_n(q - n_n) \right) . \]  

(27)
Where \( p \) and \( q \) are two independent points. By using Equations (23) and (25), Equation (27) is given by
\[
\langle P(p)P(q) \rangle = \frac{1}{N^2} \left\{ \sum_m \int dS_m P_m(p - m) \delta^2(m - m_t) \right\} \\
\times \int dS_n P_n(q - n) \delta^2(n - n_t). \tag{28}
\]

Then the three assumptions are made as follows:

1. The posterior is assumed to be uncorrelated with the true position of GW events, so we can write
\[
\langle P_l(p - m)P_n(q - n) \rangle = \left\langle P_l(p - m) \right\rangle \left\langle P_n(q - n) \right\rangle. \tag{29}
\]

2. Since \( P_l \) and \( P_n \) are posterior probability distributions estimated from two independent GW events:
\[
\langle P_l(p - m)P_n(q - n) \rangle = \langle P_l(p - m) \rangle \langle P_n(q - n) \rangle. \tag{30}
\]

3. The homogeneity of space is assumed (but not the isotropy of space when calculating ACF),
\[
\langle P_l(p - m) \rangle = P(p - m) \quad \text{and} \quad \langle P_n(q - n) \rangle = P(q - n). \tag{31}
\]

By using the assumptions above and Equation (20), we can deduce
\[
\langle P(p)P(q) \rangle = \frac{1}{N^2} \int dS_m \int dS_n P(p - m) \\
\times P(q - n) \times \left\{ \sum_m \delta^2(m - m_t) \right\} \\
\times \left\{ \sum_n \delta^2(n - n_t) \right\} \\
= \int dS_m \int dS_n P(p - m) \\
\times P(q - n) \left\{ P_0(m)P_0(n) \right\}. \tag{32}
\]

The density contrast in the field of probability is given by
\[
\delta'(p) = \frac{P(p)}{\bar{P}} - 1. \tag{33}
\]

By using Equations (19) and (27), the smeared ACF in this field is written as
\[
\omega(p, q) = \langle \delta'(p) \delta'(q) \rangle \\
= S^2 \langle P(p)P(q) \rangle - 1 \\
= \int_S dS_m \int_S dS_n P(p - m) \\
\times P(q - n) \omega_0(m, n), \tag{34}
\]

where \( p, q, m, \) and \( n \) are vectors of two-dimensional coordinate \((\alpha, \beta)\) space.

We can calculate the smeared ACF \( \omega(p, q) \) directly by \( \omega_0(m, n) \). \( \omega_0(m, n) \) only depends on \( \omega_{BBH}(\theta) \) due to the homogeneity of space. In a two-dimensional coordinate area \( S, m, \) and \( n \) can be, respectively, expressed as \((\alpha_1, \beta_1), (\alpha_2, \beta_2)\), and we can deduce
\[
\omega(p, q) = \langle \delta'(p) \delta'(q) \rangle \\
= S^2 \langle P(p)P(q) \rangle - 1 \\
= \int_S dS_n \int_S dS_m P(p - m) \\
\times P(q - n) \omega_0(\theta). \tag{35}
\]

\[
\cos \theta = \sin \left( \frac{\pi}{2} - \beta_1 \right) \cos (\alpha_1) \sin \left( \frac{\pi}{2} - \beta_2 \right) \\
\times \cos (\alpha_2) + \sin (\alpha_2) \sin (\frac{\pi}{2} - \beta_1) \cos (\frac{\pi}{2} - \beta_2) \\
\times \cos (\frac{\pi}{2} - \beta_2). \tag{36}
\]

where \( \omega_0(\theta) = b^2 \omega_{DM}(\theta) \). \( \omega_{DM}(\theta) \) is a dark matter ACF, which can be calculated by Equation (1). We transformed Equation (35) and defined two new variables \( a = p - m \) and \( b = q - n \):
\[
\omega_{\text{sm}}(p, q) = \int_S dS_a \int_S dS_b P(a)P(b) \omega_0(\theta), \tag{37}
\]

where \( \omega_{\text{sm}}(p, q) \) is averaged over two-dimensional coordinate area \( S \) for a given \( \theta \) to get \( \omega(\theta) \).

### 3. Implementation Method

#### 3.1. The Reference \( C_0 \)

For simplicity, we assume all the GW event covariance matrices are the same, and consider a covariance matrix of sky location posteriors of GW150914 as \( C_0 \).
\[
C_0 = C_{GW150914} = \begin{pmatrix} 985.7 & -70.4 \\ -70.4 & 46.5 \end{pmatrix}. \tag{38}
\]

where \( C_{GW150914} \) is the covariance matrix of GW150914 event in the directions of R.A. and decl. estimated by Bilby (Ashton & Khan 2020) using the data of Bilby’s posterior samples for GW150914 (Abbott et al. 2016b), which is observed by the LIGO Hanford (H1) and Livingston (L1) detectors.

Based on the discussion in Vijaykumar et al. (2020), the requirement of location errors in ascension and decl. should be within a degree and \( a \) is scaled by the sensitivity improvements of detectors. The \( S/N \) for the advanced GW detector in Livingston and Hanford Observatory (L-H) during the first
Thus, we set

\[ \rho_1^2 + \rho_2^2 = 2\rho_1^2, \]  

(39)

where we assume \( \rho_1^2 = \rho_2^2 \). For ET-2CE (Cosmic Explorer locations; one in Hanford, USA and one in LIGO-India location. Einstein Telescope location: proposed one in Europe), the noise of the detector is 30 times less than that of L-H during the first observing run (Punturo et al. 2010, Dwyer et al. 2015, L. S. Collaboration et al. 2015), the S/N is written as

\[ \rho_{\text{ET-2CE}}^2 = 900\rho_1^2 + 900\rho_2^2 + 900\rho_3^2 \]

\[ = 2700\rho_1^2, \]

(40)

where \( \rho_1, \rho_2, \) and \( \rho_3 \) are the S/N of the detectors, and we also assume \( \rho_1^2 = \rho_2^2 = \rho_3^2 \). By using Equations (39) and (40), we can derive:

\[ \frac{\rho_{\text{ET-2CE}}^2}{\rho_{\text{L-H}}^2} \approx 1500. \]

(41)

By using Equation (11), the error in location for ET-2CE is written as,

\[ \Delta\Omega_{\text{ET-2CE}} = \frac{1}{1500}\Delta\Omega_{\text{L-H}}. \]

(42)

Thus, we set \( a = 1/1500 \) for ET-2CE comparing with L-H during the first observing run. Then we calculate the black hole ACF smeared by the source-location posteriors from the perspective of theory and simulation.

### 3.2. Generation of BBHs Catalogs

To put our method to test, the code LOGNORMAL\_GALAXIES (Agrawal et al. 2017) is used to generate all-sky mock BBHs catalogs, which are produced for the log-normal distribution of the density field. With the BBHs bias factor used as input (Mukherjee et al. 2021), the simulations of BBHs are obtained using the dark matter power spectrum given by Eisenstein & Hu (1999) and the Planck-18 cosmological parameters (Barrena et al. 2018). The bias factor is set to 1.5, and the redshift is set to 0.3. In real observation, comparing the value of the bias factor for BBHs and galaxies can reveal the environments of the BBH mergers. If \( b_{\text{BBH}} = b_{\text{gal}} \), it means the BBH mergers happen in galaxies. If \( b_{\text{BBH}} \neq b_{\text{gal}} \), it means BBHs are predominantly distributed outside galaxies, which can provide information about the formation mechanism of BBHs and their origin (Vijaykumar et al. 2020). Further studies point out that if BBH mergers trace the distribution of dark matter more closely than the distribution of galaxies, it may imply that BBHs are of the primordial origin; whereas, if BBH mergers trace the distribution of stars in luminous and massive galaxies more closely than other distribution, it may imply that BBHs are of the stellar origin (Scelfo et al. 2018). Thus, the BBH catalogs could be generated and the ability to recover the bias could be shown. For simplicity, our conclusions are based on the fact that \( b_{\text{BBH}} \) is independent of redshift, which remains valid when bias is redshift-dependent. The steps are listed below as follows:

1. For the purpose that there are sufficient events, and the ACF does not change significantly, a 350 km\(^{-1}\) Mpc thick spherical shell is chosen. \( N \) BBHs are selected from this shell as the actual location of each GW event. Moreover, the posterior distributions of the GW events assumed Gaussian posteriors with the error matrix \( C_\ell \). The extent \( \Delta z \) of the redshift bin corresponding to this shell thickness at redshift \( z = 0.3 \) turns out to be 0.13.

2. To accurately simulate the locations of BBHs, one point is selected from each of the \( N \) posteriors. Then, the ACF is estimated by the LS estimator. The process is repeated 10,000 times. \( \omega_{\text{BBH}} \) is calculated by taking the average.

3. 1000 BBH catalogs are generated to estimate the variance. First, \( 50 \) BBH catalogs are generated as realizations of the matter field, which gives reasons for cosmic matter variance. From each of the \( 50 \) BBH catalogs, we randomly sample \( N \) points as a sub-catalog and repeat this process 20 times so that we can get 20 sub-catalogs, which may account for the fluctuations due to the sample. Thus, there are 1000 BBH sub-catalogs. \( \omega_{\text{BBH}} \) is estimated using the steps described above, that is, using one sub-catalog as the simulation of the actual GW events. Other sub-catalogs are used to estimate error bars on \( \omega_{\text{BBH}} \) by the scatter of each \( \omega_{\text{BBH}} \) calculated from 999 sub-catalogs.

4. By comparing the recovered BBH ACF \( \omega_{\text{BBH}} \) (which is calculated using the LS estimator) with the smeared theoretical BBH ACF (see Section 2.4) using the least square fitting, the bias factor \( b_{\text{BBH}} \) can be estimated.

On the one hand, when the number of GW events is inadequate, it does not reach the statistical effect, and our method is invalidated. On the other hand, when we calculate using the LS estimator, it only needs two-dimensional position coordinates for an ACF, while it needs three-dimensional position coordinates for a two-point correlation function, which means that the ACF does not fully utilize the position information of GW events as a two-point correlation function. The result in Vijaykumar et al. (2020, Figure 4) shows that the bias factor can be recovered and the two-point correlation function can achieve an appropriate statistical effect using 5000 GW events. Therefore, we also tested whether the ACF could also achieve an appropriate statistical effect using 5000 GW events to test the feasibility of our method and made a rough comparison between our method and that of Vijaykumar et al. (2020). Figure 2 shows the smeared ACF contrasted to the ACF calculated by the LS estimator from a simulation using
5000 all-sky GW events in a $350\,h^{-1}$ Mpc thick spherical shell, which is simulated by setting the redshift to 0.3 and the bias to 1.5. We found that our method is as feasible as the method of Vijaykumar et al. (2020) using 5000 GW events.

### 4. Results and Discussion

In Section 2, we mentioned that errors in ascension and decl. must be within a degree to ensure the validity of this method. We checked whether this demand could be met using ET-2CE detectors. For simplicity, we used the results of Vijaykumar et al. (2020, Figure 3), which showed the number of expected BBH mergers at various redshifts for a year of observations along with the fraction of events that are expected to be localized adequately for this type of study. The BBHs all-sky catalogs are generated using the same method in Section 3.2. The observing time solely decides the number of GW events localized adequately for ET-2CE detectors with an observational time of 3, 5 and 10 yr at redshift $z=0.3$. We quantitatively compared our results with the results of Vijaykumar et al. (2020). Figure 3 shows the ability to recover the bias factor using GW observations of BBHs in 3, 5, and 10 yr.

The bias can be recovered to within $\sim52\%$ with an observational time of three years at redshift 0.3. For an observational time of five years, the bias can be recovered to within $\sim51\%$. We can recover the bias factor to within $\sim33\%$ with an observational time of ten years. With the increase in the observational time, the number of GW events increases, and the error bar further decreases. The errors in the bias recovery using our method are larger than that using the method of Vijaykumar et al. (2020). On the one hand, it might result from the angular correlation, which does not fully utilize the position information of GW events. On the other hand, it might result from the different assumptions of the GW source-location posteriors and different values of the error matrices of GW events. The specific reasons need to be further studied in the future.

To explore the ability to probe large-scale structures using BBH observational data, we recover the bias factor by the ACF. First, we showed the feasibility of our method using 5000 BBH events and quantitatively analyzed the recovery effect of the bias factor using GW observations of BBHs in 3, 5, and 10 yr by the third-generation GW detectors and compared it with that of Vijaykumar et al. (2020), which reveals the potential of GW detection as the probe in large scale structure in the future. It reveals the environment of the BBH mergers. Although the error of the estimated bias is still larger than that of the galaxy bias due to the large location error of GW events, using GW detections to construct large-scale structures is an important method to probe dark matter distribution. In future research, to expand our work, we can completely abandon the hypothesis that the position posterior distribution of GW events is a Gaussian distribution. In addition, by using the real position posterior distribution of GW events to calculate the ACF and the two-point correlation function, we can compare their ability to probe large-scale structures.

**Acknowledgments**

We are very grateful to the anonymous referee for her/his valuable comments and suggestions. X. Shao is grateful to A. Vijaykumar for the statistical help and thanks ZhengXiang Li and Huan Zhou for their contribution to the bias factor. This
work was supported by the National Natural Science Foundation of China (grant Nos. 11922303, 119201003 and 12021003. X. Fan is supported by Hubei province Natural Science Fund for the Distinguished Young Scholars (No. 2019CFA052). Z. Cao is supported by CAS Project for Young Scientists in Basic Research YSBR-006.

**ORCID iDs**

Xilong Fan https://orcid.org/0000-0002-8174-0128

References

Aasi, J., Abbott, B., Abbott, R., et al. 2015, CQGra, 32, 074001
Abbott, B., Abbott, R., Abbott, T., et al. 2019a, ApJL, 882, L24
Abbott, B., Abbott, R., Abbott, T., et al. 2019b, PhRvX, 9, 031040
Abbott, B., Abbott, R., Abbott, T., et al. 2019c, ApJ, 883, 149
Abbott, B. P., Abbott, R., Abbott, T., et al. 2016a, PhRvX, 6, 041015
Abbott, B. P., Abbott, R., Abbott, T., et al. 2016b, PhRvL, 116, 131103
Abbott, B. P., Abbott, R., Abbott, T., et al. 2016c, PhRvL, 116, 061102
Abbott, B. P., Abbott, R., Abbott, T. D., et al. 2017, arXiv:1710.05835
Abbott, R., Abbott, T., Abraham, S., et al. 2020a, PhRvD, 102, 043015
Abbott, R., Abbott, T., Abraham, S., et al. 2020b, PhRvL, 125, 101102
Abbott, R., Abbott, T., Abraham, S., et al. 2020c, ApJL, 896, L44
Abbott, R., Abbott, T., Abraham, S., et al. 2021a, PhRvX, 11, 021053
Abbott, R., Abbott, T., Abraham, S., et al. 2021b, ApJL, 913, L7
Acernese, F. A., Agathos, M., Agatsuma, K., et al. 2014, CQGra, 32, 024001
Adhikari, R. X., Ajith, P., Chen, Y., et al. 2019, CQGra, 36, 245010
Aghanim, N., Douspis, M., Hurier, G., et al. 2019, A&A, 623, A47
Agrawal, A., Makiya, R., Chiang, C.-T., et al. 2017, JCAP, 2017, 003
Ashton, G., Ackley, K., Hernandez, I. M., & Piotrzkowski, B. 2020, arXiv:2009.12346
Ashton, G., & Khan, S. 2020, PhRvD, 101, 064037
Aso, Y., Michimura, Y., Soniya, K., et al. 2013, PhRvD, 88, 043007
Banagiri, S., Mandic, V., Scarlata, C., & Yang, K. Z. 2020, PhRvD, 102, 063007
Barrena, R., Srebryanska, A., Ferragamo, A., et al. 2018, A&A, 616, A42
Calore, F., Cuoco, A., Regimbau, T., Sachdev, S., & Serpico, P. D. 2020, PhRvR, 2, 023314
Camera, S., & Nishizawa, A. 2013, PhRvL, 110, 151103
Chen, H.-Y., Haster, C.-J., Vitale, S., Farr, W. M., & Isi, M. 2020, arXiv:2009.14057
Dwyer, S., Sigg, D., Ballmer, S. W., et al. 2015, PhRvD, 91, 082001
Eisenstein, D. J., & Hu, W. 1999, ApJ, 511, 5
Feeney, S. M., Peiris, H. V., Williamson, A. R., et al. 2019, PhRvL, 122, 061105
Finn, L. S., & Chernoff, D. F. 1993, PhRvD, 47, 2198
Hu, Q., Li, M., Niu, R., & Zhao, W. 2021, PhRvD, 103, 064057
Iyer, B., Souradeep, T., Unnikrishnan, C., et al. 2011, Report No. LIGO-M1100296, Indian Initiative in Gravitational-wave Observations
Landy, S. D., & Szalay, A. S. 1993, ApJ, 412, 64
L. S. Collaboration 2019, LIGO Document: LIGO, 1900409
L. S. Collaboration, et al. 2015, LIGO Document: LIGO, 1400316
Mukherjee, S., Wandelt, B. D., Nissanke, S. M., & Silvestri, A. 2021, PhRvD, 103, 043520
Namikawa, T. 2021, JCAP, 2021, 036
Namikawa, T., Nishizawa, A., & Taruya, A. 2016a, PhRvL, 116, 121302
Namikawa, T., Nishizawa, A., & Taruya, A. 2016b, PhRvD, 94, 024013
Payne, E., Banagiri, S., Lasky, P. D., & Thrane, E. 2020, PhRvD, 102, 102004
Punuto, M., Abernathy, M., Acernese, F., et al. 2010, CQGra, 27, 194002
Raccanelli, A., Kovetz, E. D., Bird, S., Cholis, I., & Munoz, J. B. 2016, PhRvD, 94, 023516
Ruan, W.-H., Liu, C., Guo, Z.-K., Wu, Y.-L., & Cai, R.-G. 2020, NatAs, 4, 108
Scelfo, G., Bellomo, N., Raccanelli, A., Matarrese, S., & Verde, L. 2018, JCAP, 2018, 039
Schutz, B. F. 1986, Natur, 323, 310
Simon, P. 2007, A&A, 473, 711
Sinha, M., & Garrison, L. H. 2020, MNRAS, 491, 3022
Vijaykumar, A., Saketh, M., Kumar, S., Ajith, P., & Choudhury, T. R. 2020, arXiv:2005.01111
Zhang, C., Gong, Y., Liu, H., et al. 2021, PhRvD, 103, 103013