# Influence of higher-order waveform multipoles for the detection of eccentric binary black hole mergers 

Adam Rebei<br>NCSA Gravity Group<br>University of Illinois at Urbana-Champaign

Eliu Huerta, Roland Haas, Sibo Wang, Daniel Johnson, Daniel George

## Outline

$>$ Higher Order Modes
$>$ Past Studies
$>$ Motivations
> Gravitational Waveform Results
$>$ Signal-to-Noise Ratio Results

## Higher Order Modes

$$
h(t)=h_{+}-i h_{\times}=\sum_{l=2}^{\infty} \sum_{m=-l}^{l} h^{l m} Y_{-2}^{l m}(\theta, \varphi)
$$

## 

Python Open Source Waveform Extractor (POWER): An open source, Python package to monitor and post-process numerical relativity simulations (Johnson, Huerta, Haas Class. Quantum Grav. Volume 35, 2018)

## Past Studies

## Focus on quasi-circular binary black hole systems that are both spinning and non-spinning

> Highest fitting factors are observed for equal mass systems, since the amplitude of the higher order modes is around 2 magnitudes smaller than the amplitude of the dominant mode, $(2,2)$.
$>$ As the mass ratio increases, the relative amplitude of the higher order modes increase and the fitting factor decreases (Pan et. al Phys. Rev. D 84, 2011)
$>3 \mathrm{M} \leq \mathrm{m} 1, \mathrm{~m} 2 \leq 25 \mathrm{M}$, For binaries with high mass-ratios and inclination $0.31 \leq \theta \leq 2.68$, including higher order modes could increase the signal-to-noise ratio by as much as $8 \%$ (Brown, Kumar, and Nitz Phys. Rev. D 87, 2013)

## Motivations



Detection of black holes in M22 (Strader et al, Nature, 2012)
$>$ Gravitational-wave observations of binary black holes currently rely on theoretical models that use only the dominant multipoles ( $\mathrm{l}=2,|\mathrm{~m}|=2$ ) to predict the gravitational radiation during inspiral, merger and ringdown.
$>$ No study in the literature that has shed light on the importance of higher-order modes in the context of eccentric binary mergers
> Merger rate of eccentric black hole mergers underpredicted by a factor of 100 (previous predicted rate: $5 \mathrm{Gpc}^{-3} \mathrm{yr}^{-1}$ ) (Samsing, Askar, and Giersz ApJ, 2018)
> C Rodriguez corrects previous rate estimates for LIGO's eccentric mergers (Rodriguez et al, 2017)
> State-of-the-art waveform model for eccentric black hole mergers. Recent detections may have eccentricity content (Huerta et al, PRD, 2018)

Rebei et al, in preparation

## Equal Mass Binaries

$$
(\ell,|m|)=\{(2,2),(3,3),(4,4),(2,1),(3,2)\}
$$

## Low Eccentricity

High Eccentricity




## q=8 Black Hole Binaries

$$
(\ell,|m|)=\{(2,2),(3,3),(4,4),(2,1),(3,2)\}
$$

Low Eccentricity

High Eccentricity


Effects of Eccentricity
Rebei et al, in preparation
$q=1 ; e=0.1$




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Rebei et al, in preparation

## Signal-to-Noise Ratio

$$
\mathrm{q}=1 ; \mathrm{e}=0.45, \text { total mass }=100 \mathrm{M}_{\odot}
$$



Using Numerical Relativity Waveforms at least 1500M long


## Summary

$>$ Higher order modes in higher mass-ratio binaries $\rightarrow$ Access to wider range of $(\theta, \varphi)$ combinations
$>$ Including higher order modes isn't always better than using only the $(2,2)$ mode
$>$ SNR can increase by as much as $20 \%$ when higher order modes are used for asymmetric mass-ratio binaries
$>$ In the future:
$>$ Finish simulations with mass-ratio 10 to complete analysis
$>$ Explore spinning and eccentric binaries

## NCSA Gravity Group at the 2018 APS April Meeting

1.Ed Seidel: Numerical Relativity 1980-2000s: The era of sharpening our tools and exploring Einstein's physics
2.Daniel George: Deep Learning for Real-time Gravitational Wave Detection and Parameter Estimation: Results with Advanced LIGO Data
3.Adam Rebei: Influence of higher-order waveform multipoles for the detection of eccentric binary black hole mergers
4.Roland Haas: Assessing confidence in numerical relativity waveforms of binary neutron star mergers 5.Hongyu Shen: Glitch Classification and Clustering for LIGO with Deep Transfer Learning (poster)
6.Eliu Huerta: Detection and characterization of eccentric compact binary coalescence at the interface of numerical relativity, analytical relativity and machine learning
7.Hongyu Shen: Denoising Gravitational Waves using Deep Learning with Recurrent Denoising Autoencoders 8.Roland Haas: BOSS-LDG using Blue Waters for LIGO data analysis (poster)
9.Vedant Puri: Scheduled Relaxation Jacobi Method for Initial Data Problems
10.Shawn Rosofsky: Study of f-mode Oscillations in Numerical Relativity Simulations of Perturbed Neutron Stars and Highly Eccentric Binary Neutron Star Mergers
11.Pablo Brubeck: On the Schur complement of the nearest Kronecker product preconditioner for elliptic boundary value problems
12.Haris Markakis: Helmholtz's third theorem in numerical general relativity
13.Miguel Holgado: Pulsar Timing Constraints on the Fermi Massive Black-Hole Binary Blazar Population

