

Using astrophysical knowledge in gravitational-wave data analysis of binary inspirals

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Outline

1 Introduction

- Gravitational waves
- LIGO/Virgo



2 GW parameter estimation

- Signal and noise
- Markov-chain Monte Carlo algorithm
- Example SPINSPIRAL analysis
- MCMC examples
- Analysis of a BH-NS signal
- Analysis of a BH-BH signal
- The importance of having spins

3 Using astrophysical information

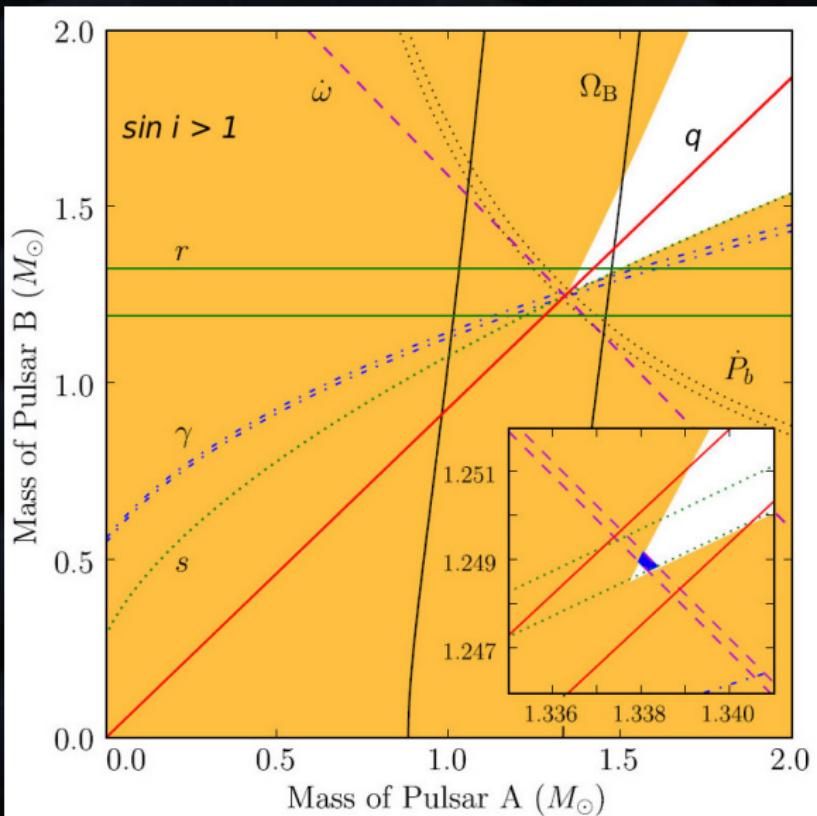
- Example: GRB without spin
- Example: GRB with spin

4 Conclusions

Gravitational waves

GWs:

- “Ripples in spacetime”
- Predicted by Einstein’s theory of General Relativity
- *Indirectly* observed for the Hulse-Taylor binary pulsar:



Gravitational waves

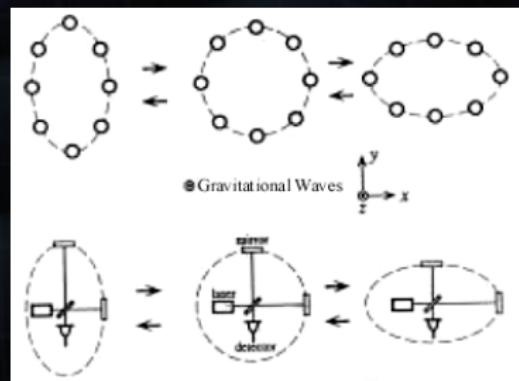
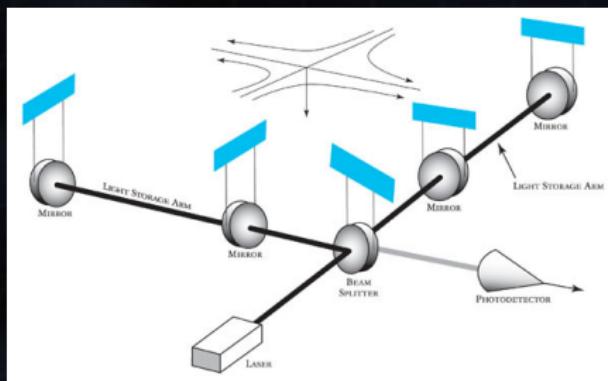
Gravitational waves...

- propagate transversely at the speed of light
- are quadrupole radiation at the lowest order
- stretch and squeeze spacetime in two polarisations
- allow us to measure their amplitude



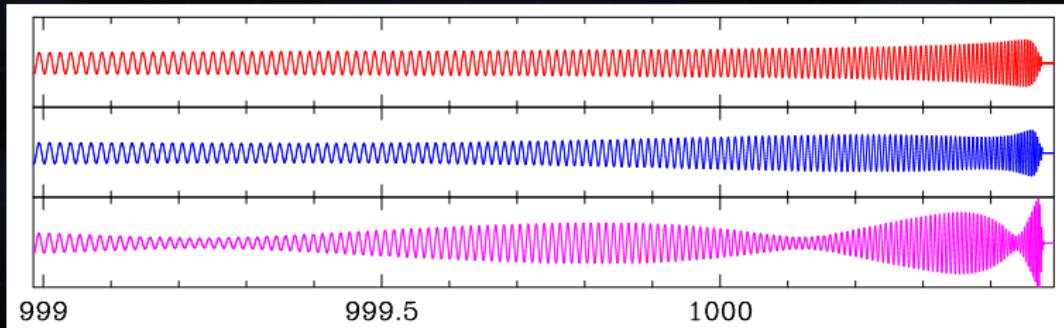
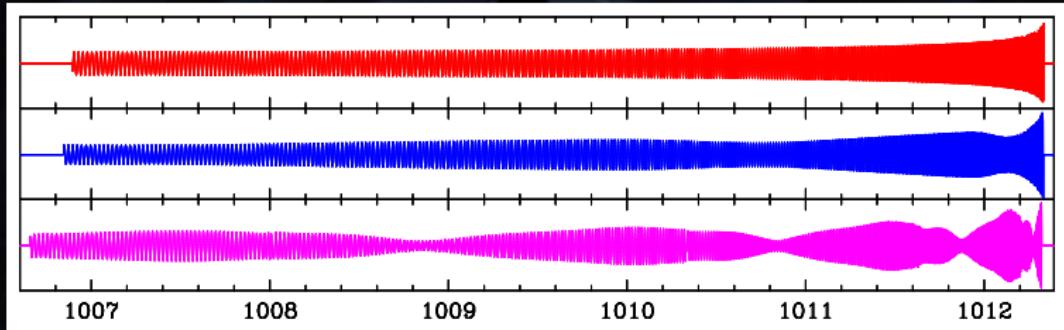
- Strain: $h(t) = h_+(t)F_+(t) + h_\times(t)F_\times(t) = \frac{\delta L(t)}{L} \sim 10^{-22}$

Laser Interferometer GW Observatory (LIGO)



Inspiral waveforms with increasing spin

LIGO and Virgo detect the last ~ 10 s of a binary inspiral:

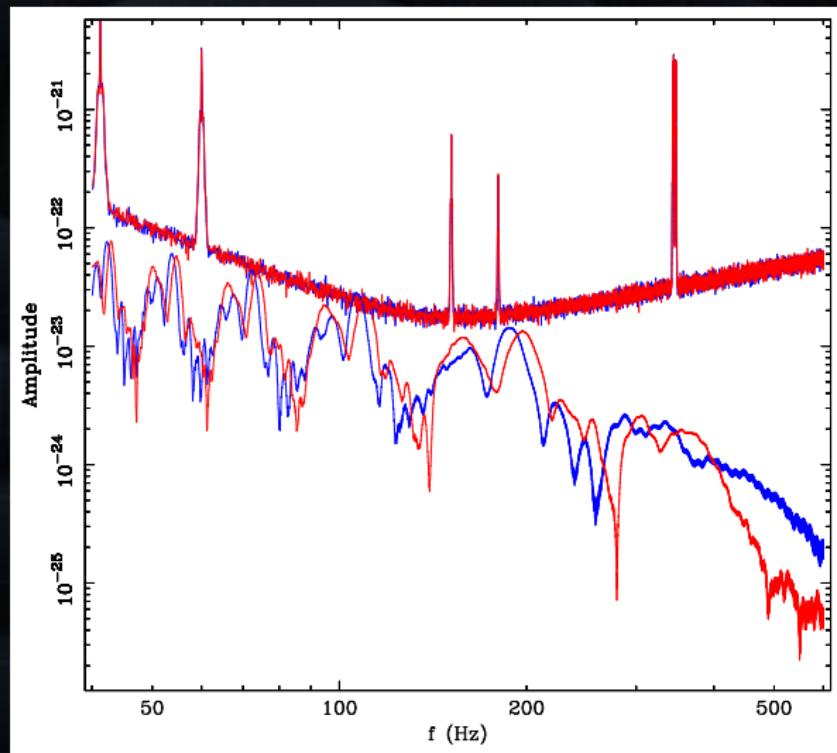


$$a_{\text{spin}} \equiv S/M^2 = 0.0, 0.1 \text{ and } 0.5$$

Signal injection into detector noise

Example:

- Using two 4-km detectors H1, L1
- Inject signal coherently
- $\Sigma \text{ SNR} = 17$
- Retrieve physical parameters using MCMC

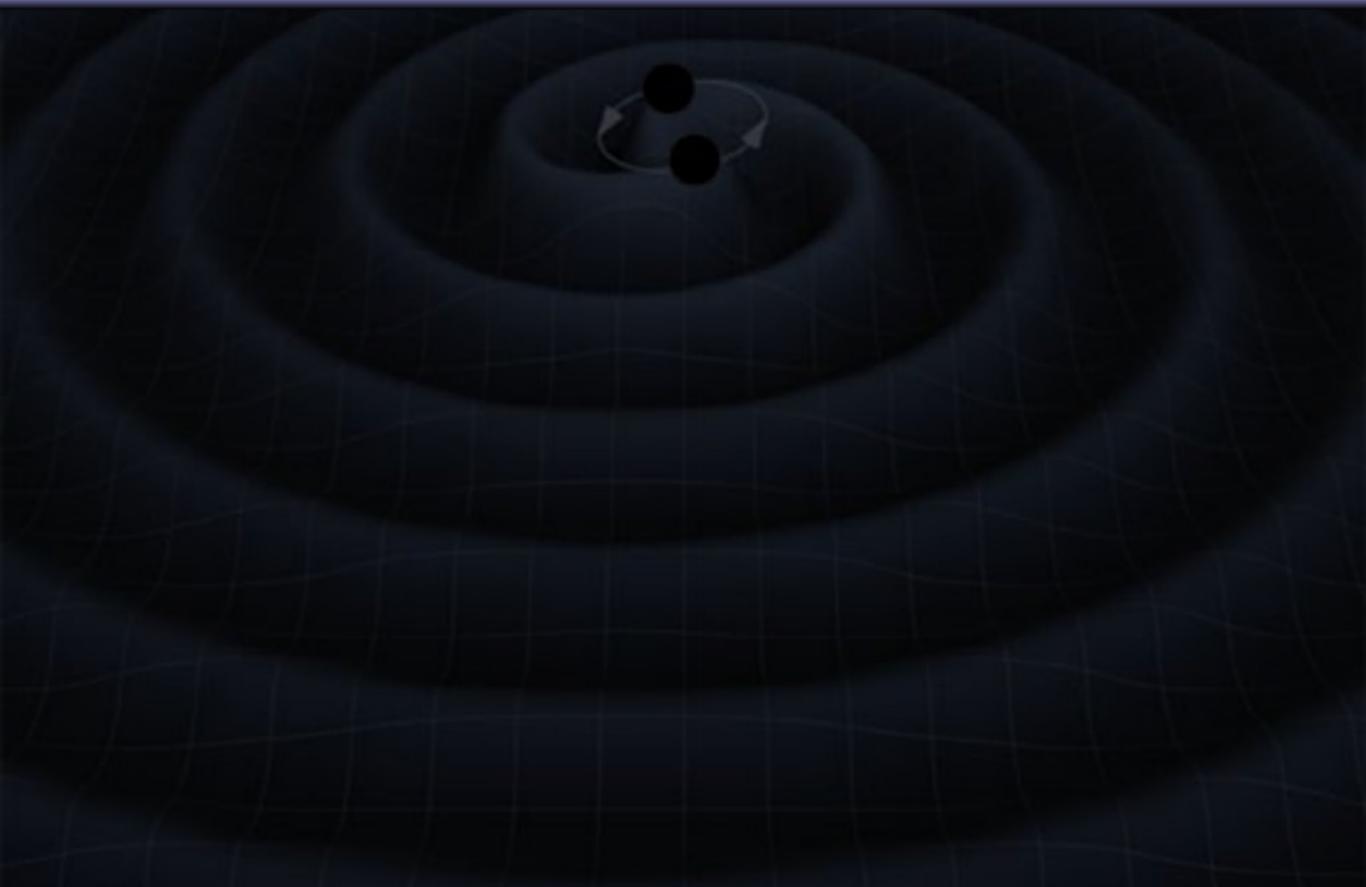


SPINSPIRAL code

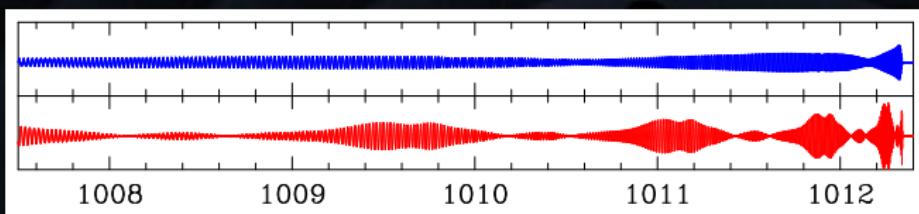


- Use Markov-Chain Monte Carlo for parameter estimation
- Follow-up after detection
- Gaussian, stationary noise or LIGO/Virgo detector data
- Analyse software injections, hardware injections, detection candidates/interesting events
- Include spin in injections and analysis
- Use any network composed of LIGO/Virgo detectors:
 - $\text{PDF}(\vec{\lambda}) \propto \text{prior}(\vec{\lambda}) \times \prod_i L_i(d|\vec{\lambda})$
- Result: posterior probability-density function (**PDF**) of the parameter set that describes the model (9–12–15 D)

SPINSPIRAL example



Correlations increase with spin

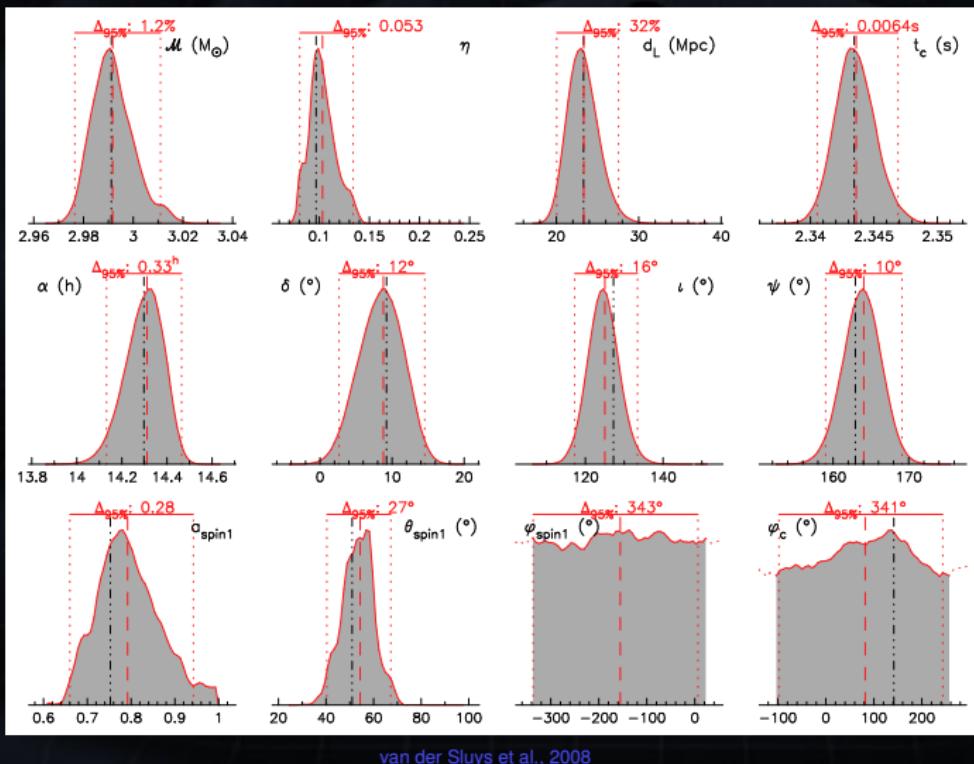


	M_c	η	a_{spin}	ϑ_{SL}	R.A.	Dec.
M_c		0.22	0.42	0.17	-0.40	0.19
η	-0.27		-0.34	-0.53	-0.07	-0.04
a_{spin}	-0.61	0.89		-0.04	0.11	0.62
ϑ_{SL}	0.66	-0.87	-0.99		0.02	-0.34
R.A.	-0.36	0.01	0.02	-0.02		0.12
Dec.	-0.23	0.08	0.18	-0.20	-0.05	

Parameters:

- BH-NS
- H1 & L1
- $M_1 = 10 M_\odot$
- $M_2 = 1.4 M_\odot$
- $a_{\text{spin}} = 0.1, 0.8$
- $\theta_{\text{SL}} = 55^\circ$
- Network SNR ≈ 25

MCMC results for the analysis of a BH-NS signal

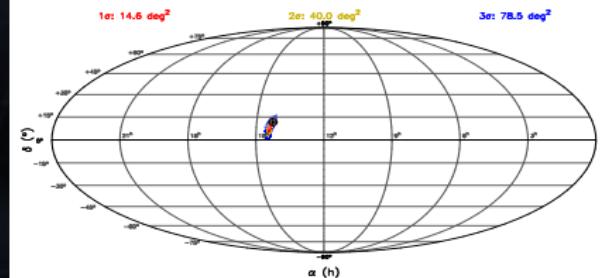
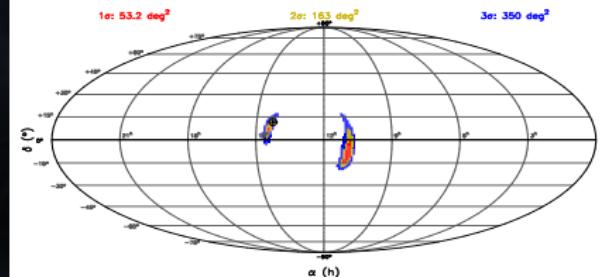
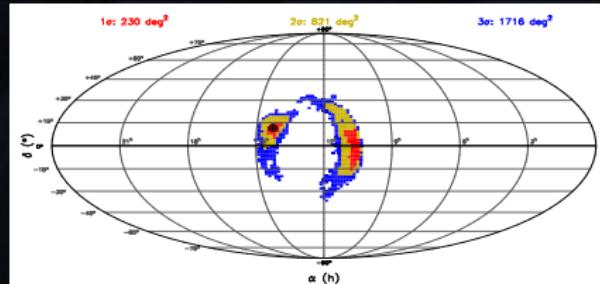


van der Sluys et al., 2008

Parameters:

- H1, L1, V
- $M = 10, 1.4 M_{\odot}$
- $d_L = 22.4 \text{ Mpc}$
- $a_{\text{spin}} = 0.8$,
 $\theta_{\text{SL}} = 55^\circ$
- $\Sigma \text{ SNR} \approx 17.0$
- simulated noise
- Black dash-dotted line: injection
- Red dashed line: median
- Δ 's: 95% probability

Sky position for signals with different spins



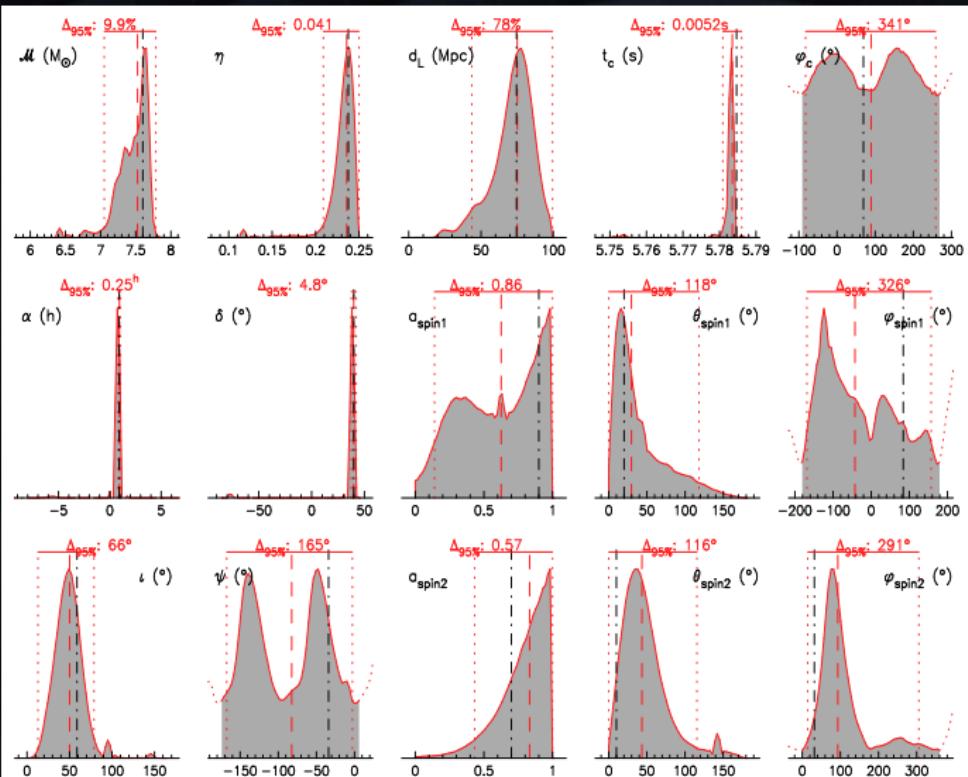
Spinning BH, non-spinning NS:
 $10 + 1.4 M_{\odot}$, 16–22 Mpc, $\Sigma \text{SNR} = 17$

2 detectors, $a_{\text{spin}} = 0.0$
2- σ accuracy: 821 $^{\circ2}$

2 detectors, $a_{\text{spin}} = 0.5$
2- σ accuracy: 163 $^{\circ2}$

3 detectors, $a_{\text{spin}} = 0.5$
2- σ accuracy: 40 $^{\circ2}$

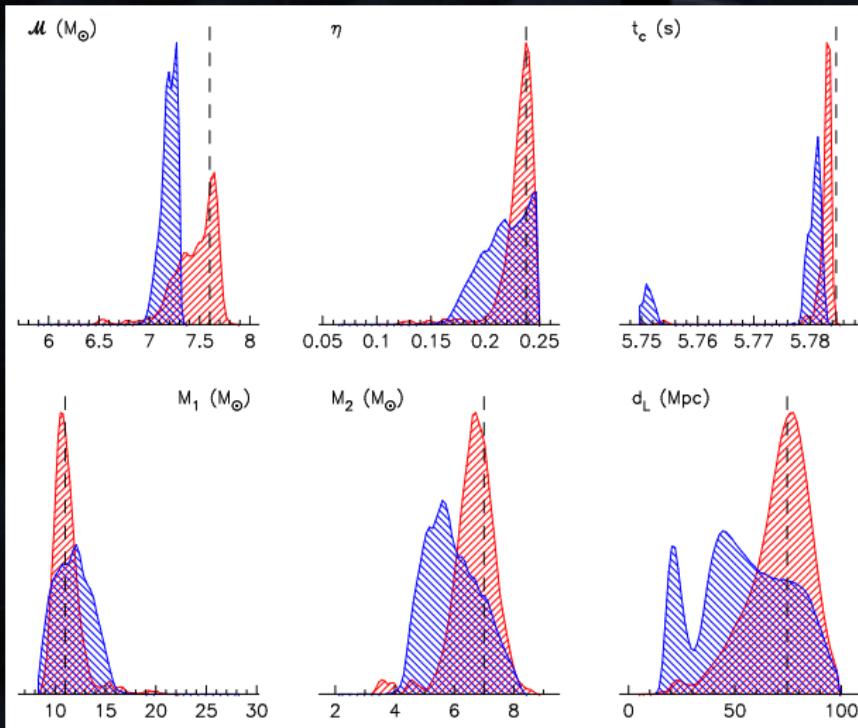
Analysis of a BH-BH signal with spins



HS-2:

- 3.5-pN waveform
- 3 detectors (H1,L1,V)
- $\mathcal{M} = 7.6 M_{\odot}$,
 $\eta = 0.238$;
 $M_1 = 11.0 M_{\odot}$,
 $M_2 = 7.0 M_{\odot}$
- $a_{s1,2} = 0.9, 0.7$
- $\theta_{s1,2} = 10, 20^{\circ}$
- $d_L = 74.5 \text{ Mpc}$
- $\Sigma \text{ SNR}=15$
- simulated noise

The importance of having spins in your analysis



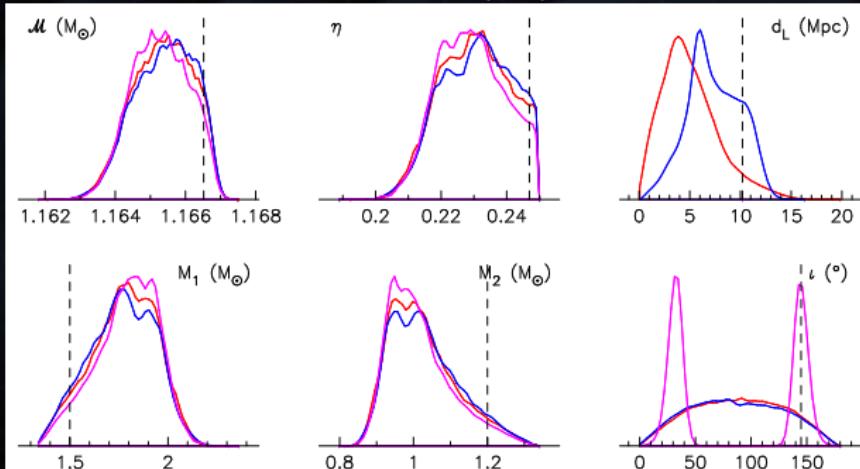
Signal with spins

Analysis with spinning template

Analysis with non-spinning template

Using astrophysical data to constrain parameters

1 detector (H1):



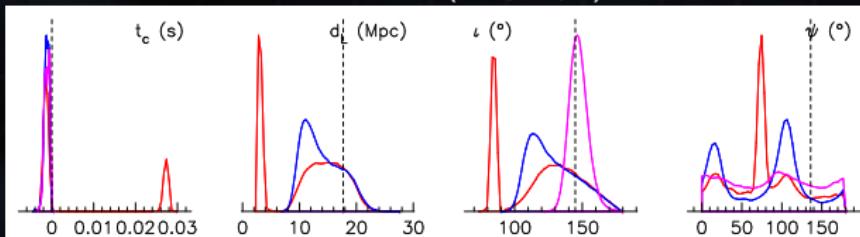
NS-NS, non-spinning:

$$1.2 + 1.5 M_\odot$$

$$d_L \approx 10.2 - 17.8 \text{ Mpc}$$
$$(\Sigma \text{ SNR}=15.0)$$

No astrophysical
information

3 detectors (H1,L1,V):



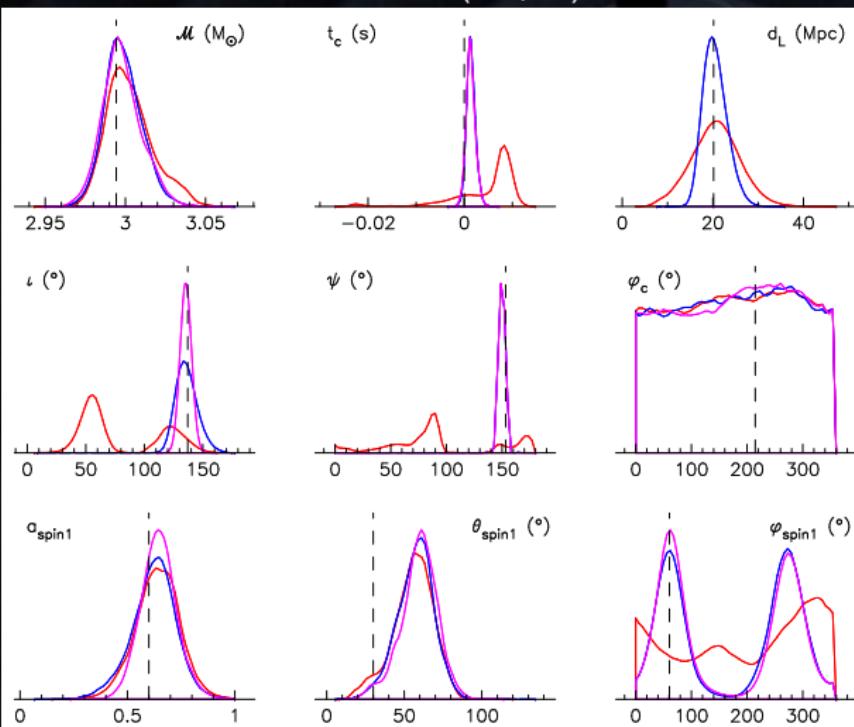
Sky position and distance
known

van der Sluys et al., in preparation

See also: Nissanke et al., 2010

Using astrophysical data to constrain parameters

2 detectors (H1,L1):



BH-NS, spinning BH:
 $10. + 1.4 M_\odot$
 $d_L \approx 20.2 \text{ Mpc}$
($\Sigma \text{ SNR}=15.0$)

No astrophysical
information

Sky position known

Sky position and distance
known

Conclusions

SPINSPIRAL

- SPINSPIRAL can recover the 12–15 parameters of a binary inspiral, including one or two spins, using an MCMC technique
 - Sky-position reconstruction ($\text{few} \times 10^{\circ}{}^2$) is poor for astrophysical standards
 - Combination of position, distance and time can lead to association with an electromagnetic detection (e.g. GRB)

Taking into account spins

- The inclusion of spin adds significantly to the number of dimensions (9–12–15) and introduces (strong) correlations
 - Failing to take into account spin can result to biases in especially mass parameters

Conclusions (numbers are preliminary)

Using astrophysical knowledge for GW data analysis: no spins

- Knowing the sky position of a source improves determination of:
 - distance ($\sim 20 - 50\%$)
 - inclination
- Knowing the position *and distance* improves inclination further, also in 1-detector analysis

Using astrophysical knowledge for GW data analysis: spins

- Knowing the sky position of a source improves determination of:
 - distance ($\sim 50\%$)
 - inclination, polarisation angle ($50 - 90\%$)
 - masses ($\sim 20\%$)
 - spin angles
- Knowing the position *and distance* improves:
 - spin magnitude ($\sim 20\%$)

End...



Predicted detection rates

Realistic estimate:

	Rates (yr⁻¹)			Horizon (Mpc)		
	NS-NS	BH-NS	BH-BH	NS-NS	BH-NS	BH-BH
Initial	0.015	0.004	0.01	32	67	160
Enhanced	0.15	0.04	0.11	71	149	349
Advanced	20	5.7	16	364	767	1850

Plausible, optimistic estimate:

	Rates (yr⁻¹)			Horizon (Mpc)		
	NS-NS	BH-NS	BH-BH	NS-NS	BH-NS	BH-BH
Initial	0.15	0.13	1.7	32	67	160
Enhanced	1.5	1.4	18	71	149	349
Advanced	200	190	2700	364	767	1850

Estimates assume $M_{\text{NS}} = 1.4 M_{\odot}$ and $M_{\text{BH}} = 10 M_{\odot}$

CBC group, rates document

MCMC analyses

MCMC parameters

Masses: $\mathcal{M} \equiv (M_1 + M_2) \eta^{3/5}$ & $\eta \equiv \frac{M_1 M_2}{(M_1 + M_2)^2}$, distance: $\log d_L$, time and phase at coalescence: t_c & φ_c , position: α & $\sin \delta$, spin magnitude: $a_{\text{spin}_{1,2}}$, spin orientation: $\cos \theta_{\text{spin}_{1,2}}$ & $\varphi_{\text{spin}_{1,2}}$ & binary orientation: $\cos(\iota)$ & ψ

MCMC set-up

- ≥ 5 serial chains per run, starting from offset parameter values
- Chain length: $\sim \text{few} \times 10^6$ states; burn-in: $\sim \text{few} \times 10^5$ states
- Run time: 10 days on a 2.8 GHz CPU for 1.5-pN waveform;
 $\sim 2.5 \times$ longer for 3.5-pN

MCMC analyses

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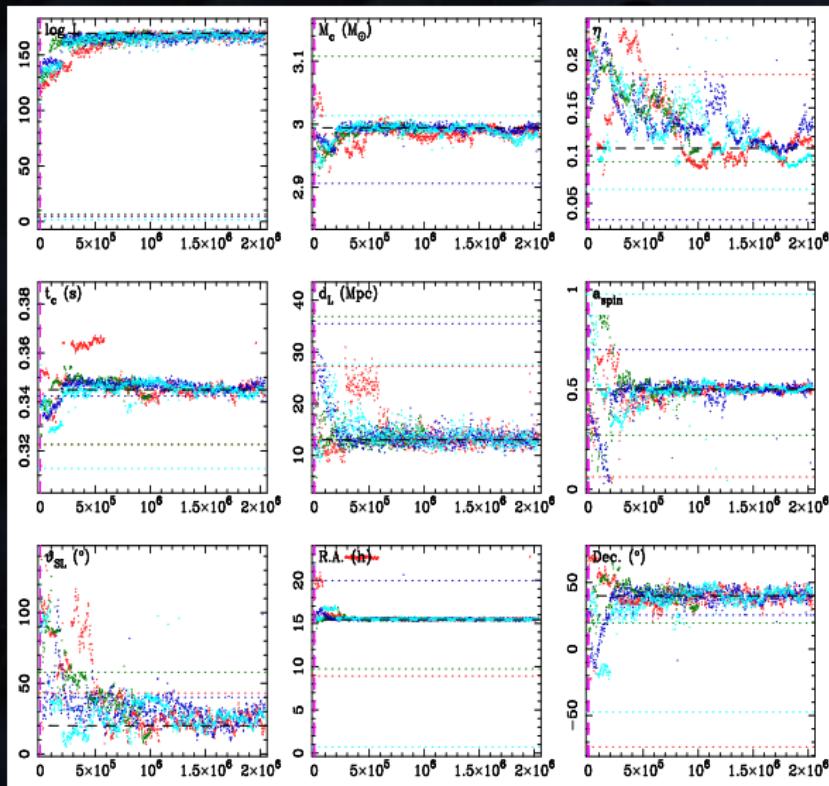
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Analysis details: BH-NS signal

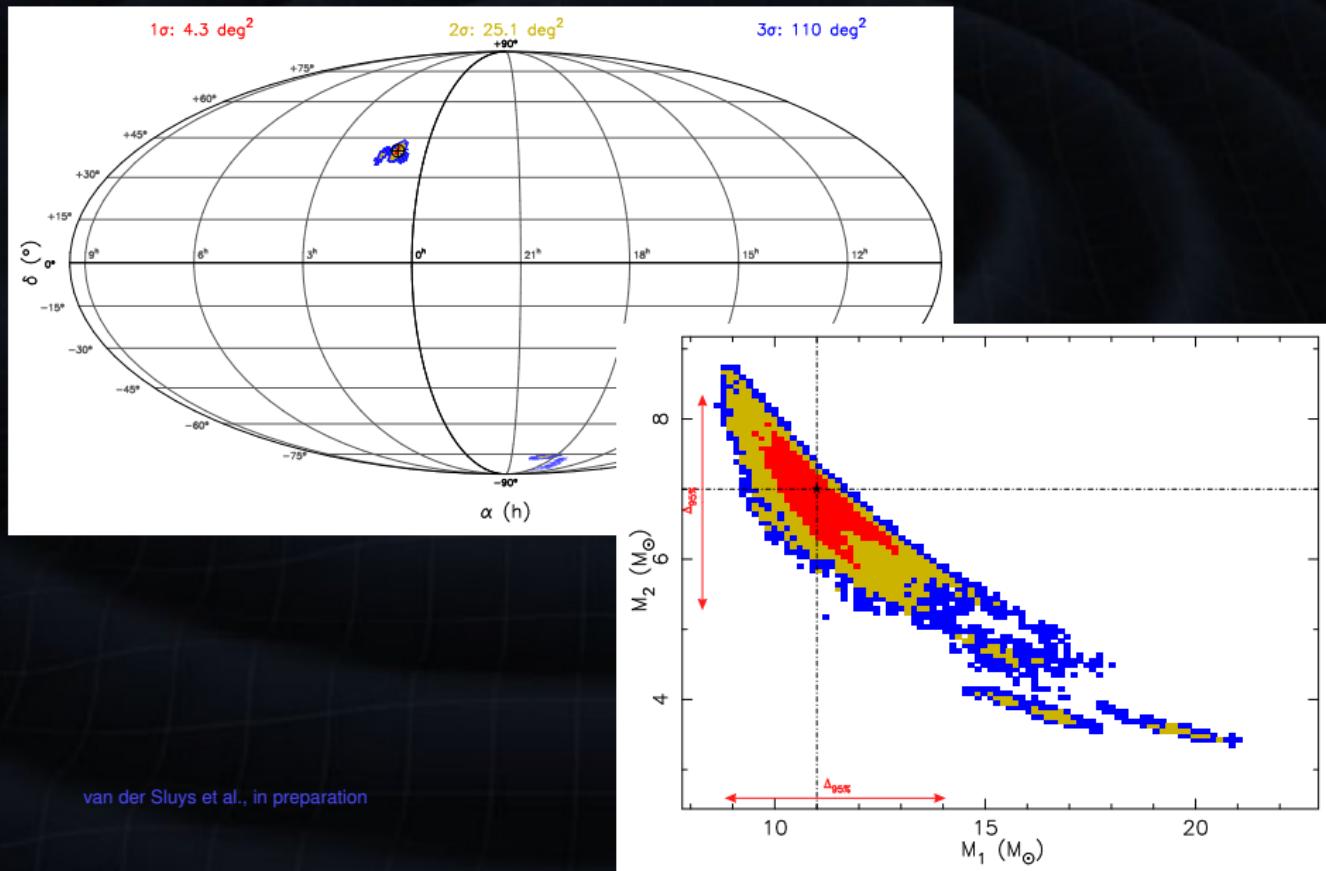
- Signals injected in simulated noise for H1L1V @ SNR ≈ 17.0
- Fiducial binary: $M_{1,2} = 10 + 1.4 M_\odot$, $d_L = 16 - 23$ Mpc
- Spin: $a_{\text{spin}} = 0.0, 0.1, 0.5, 0.8$, $\theta_{\text{SL}} = 20^\circ, 55^\circ$

Convergence of chains

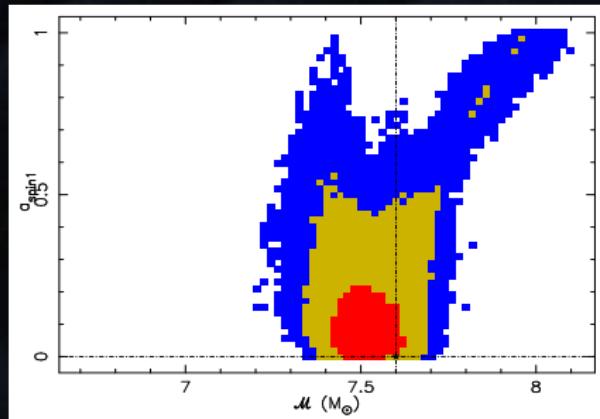


- Dots: starting values
- Dashes: injection values

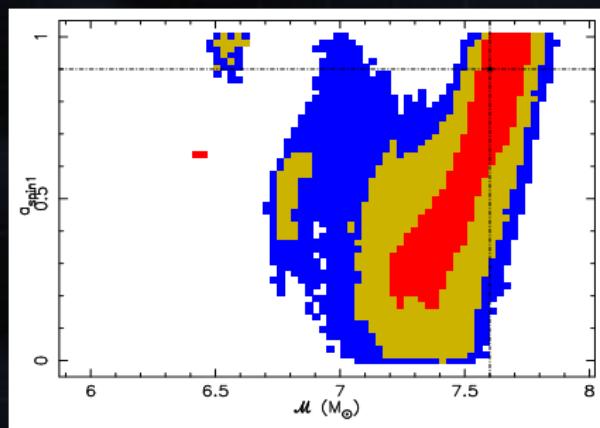
Analysis of a BH-BH signal with spins



The nuisance of having spins in your analysis



Signal **without** spins,
analysis with spinning template



Signal **with** spins,
analysis with spinning template