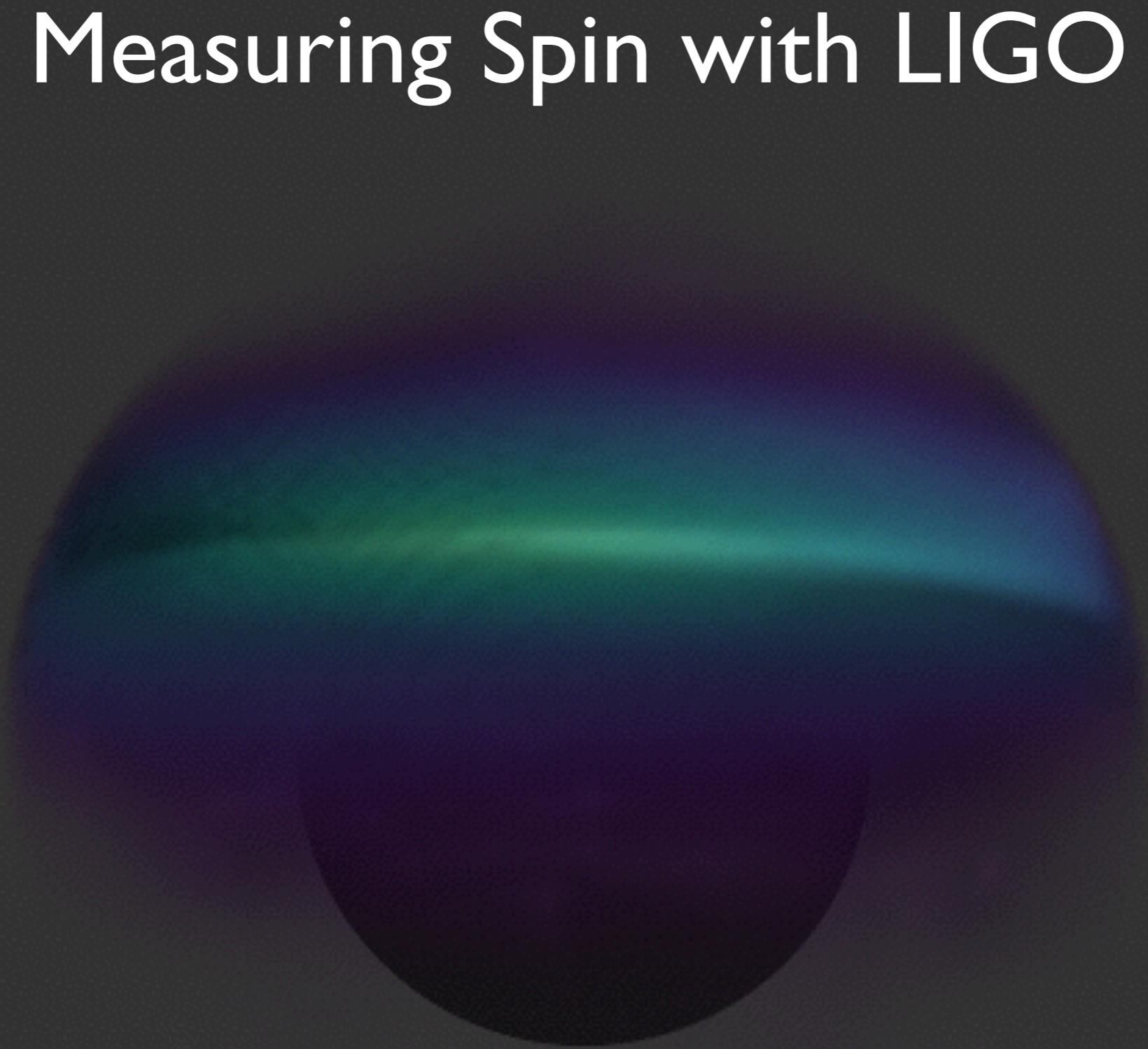


# Measuring Spin with LIGO



Ben Farr

KITP: Massive Stars Workshop



THE UNIVERSITY OF  
CHICAGO

# 15 Model Parameters

Intrinsic	Extrinsic
Masses (2)	Location (2) Distance (1) Inclination (1) Orientation (2)
Spins (6)	Merger Time (1)

# Bayesian Inference

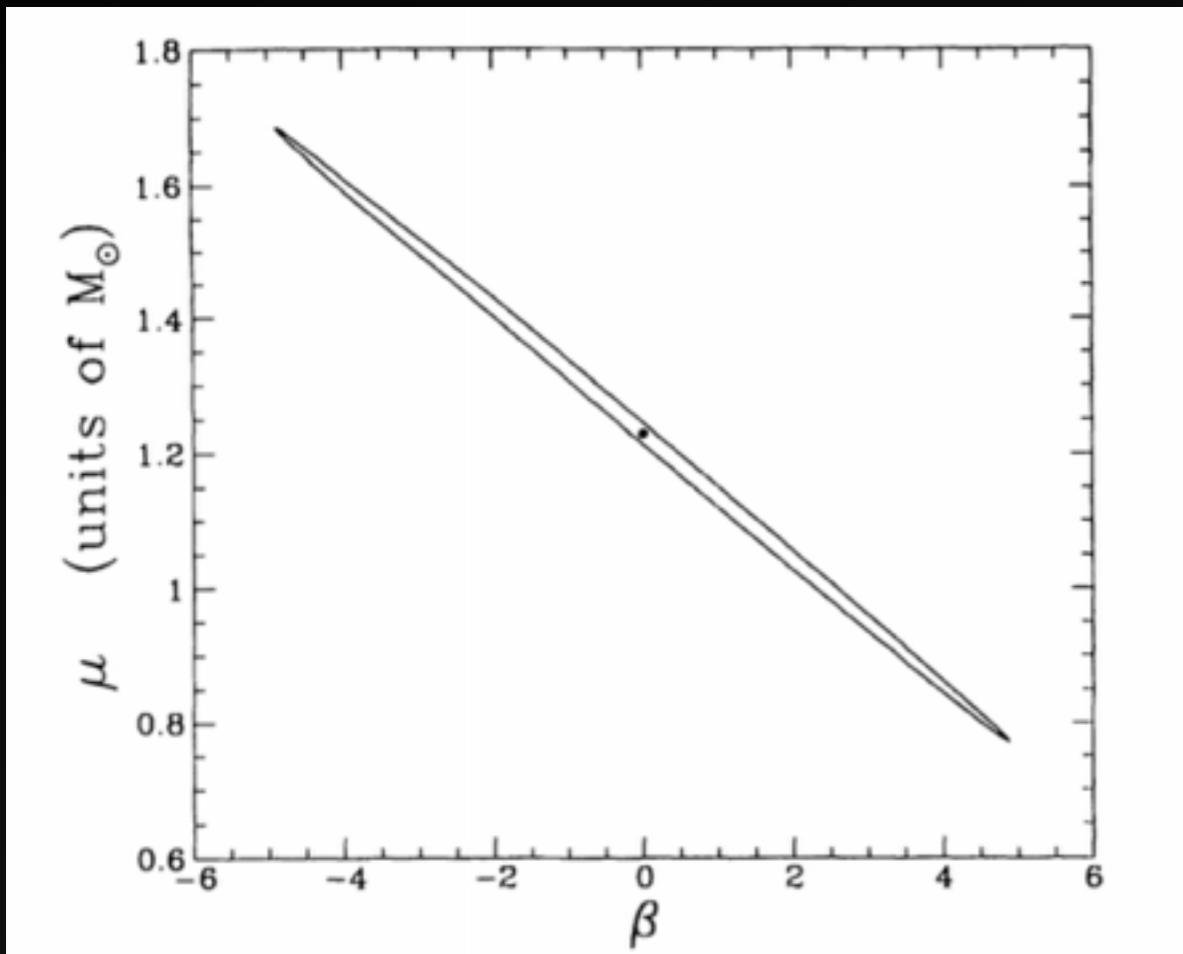
$$p(\vec{\theta}|d) \propto p(\vec{\theta})p(d|\vec{\theta})$$

Posterior

Prior

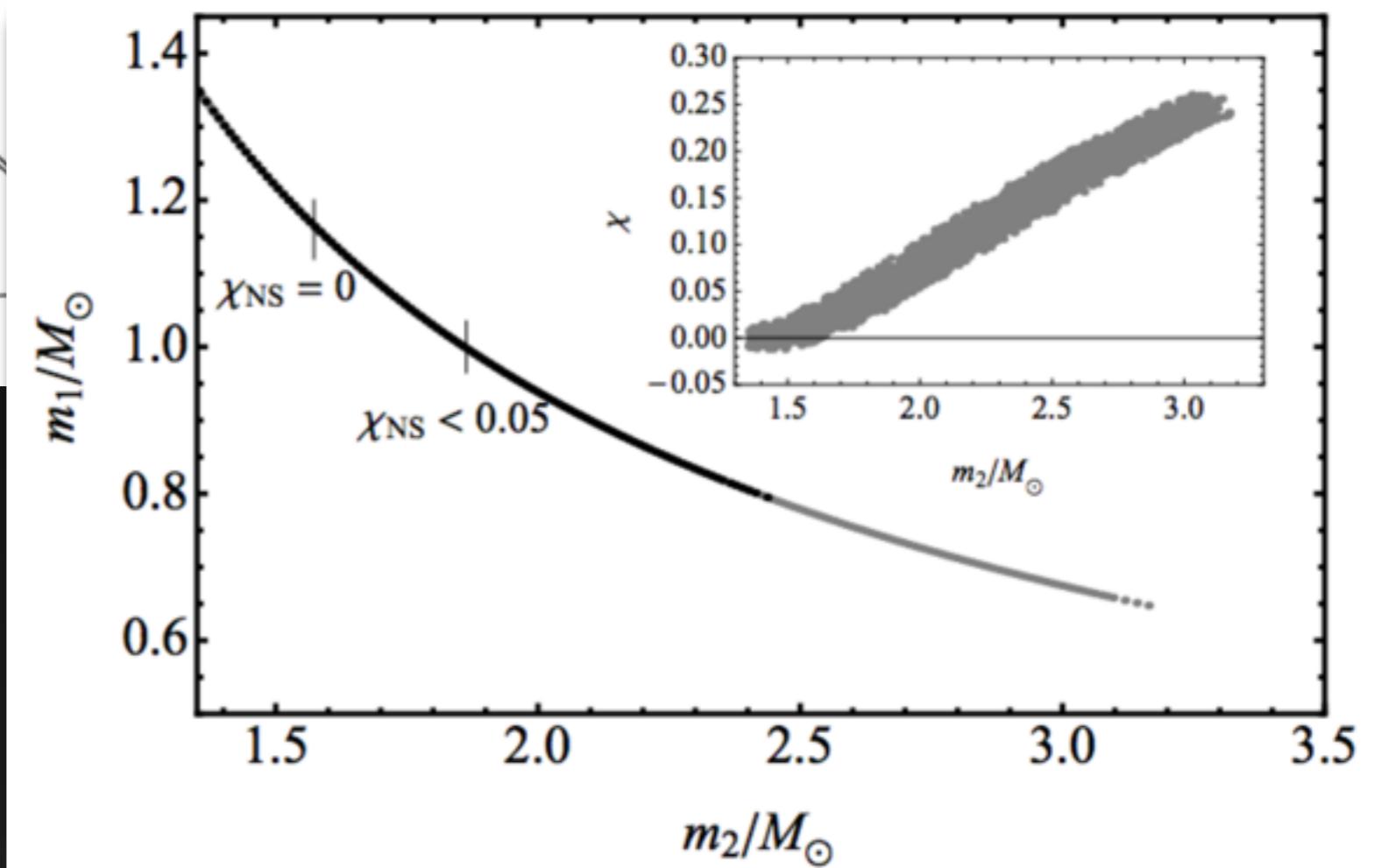
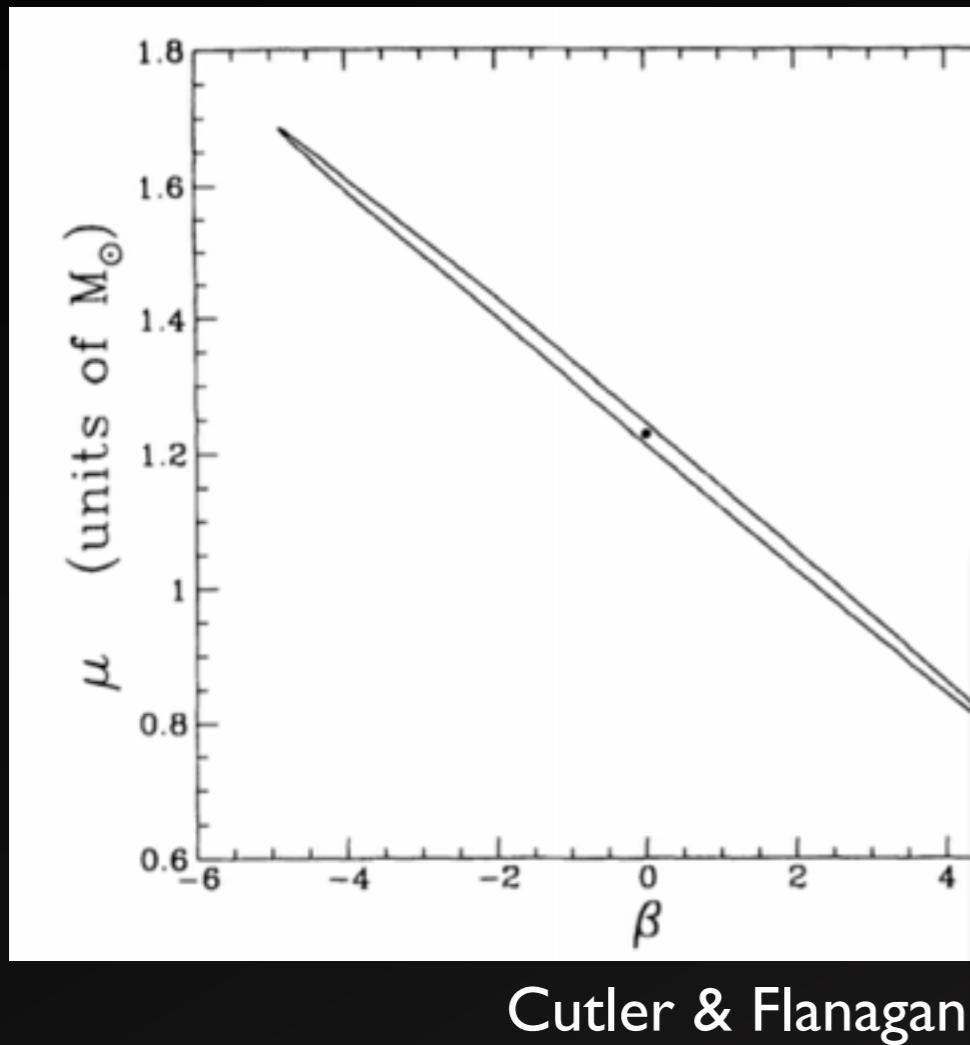
Likelihood

# Mass-Spin Degeneracy



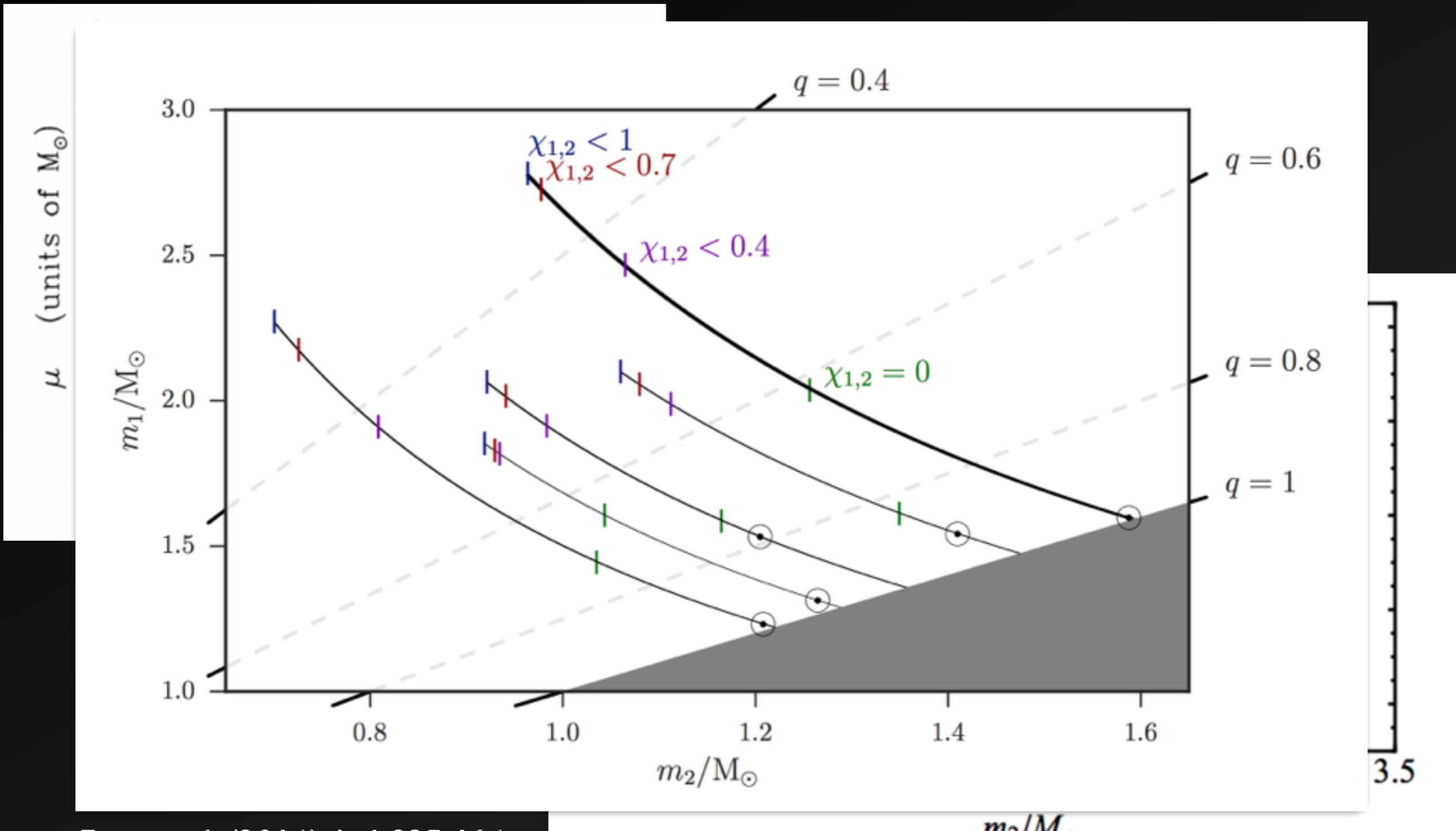
Cutler & Flanagan (1994)

# Mass-Spin Degeneracy



Hannam et al. (2013)

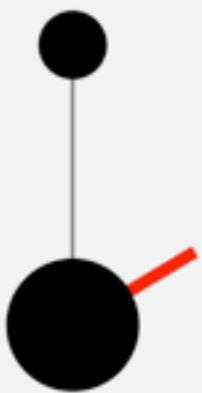
# Mass-Spin Degeneracy



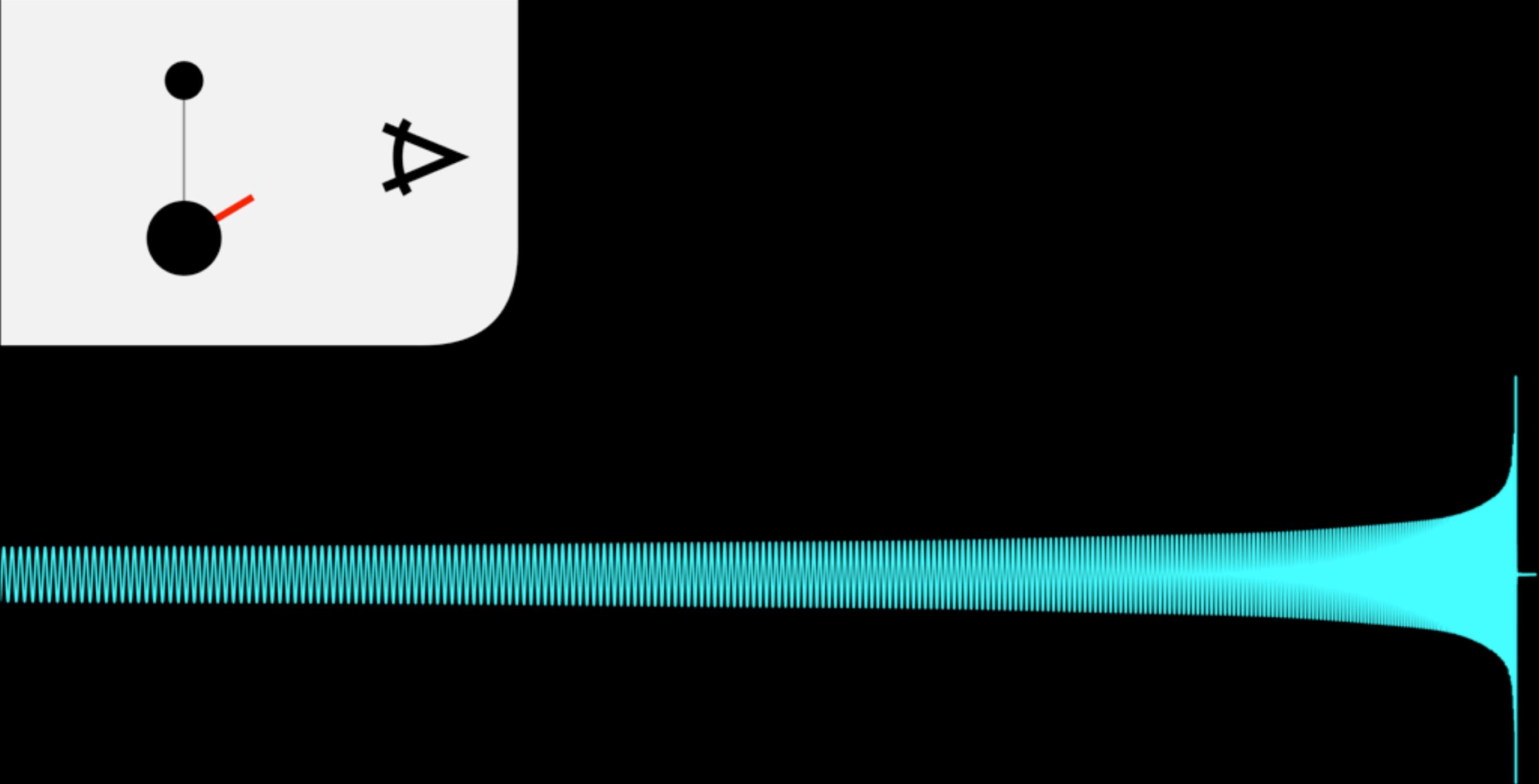
Farr et al. (2016) ApJ 825 116

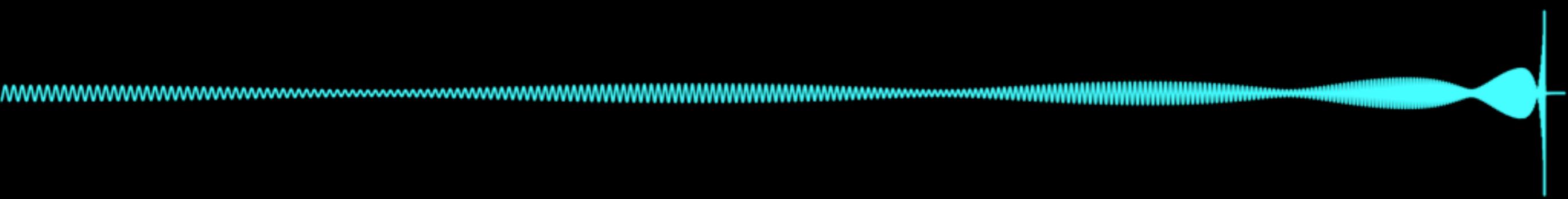
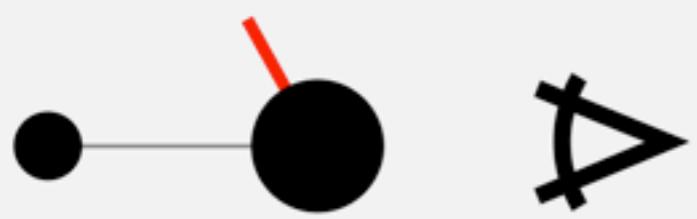
$m_2/M_\odot$

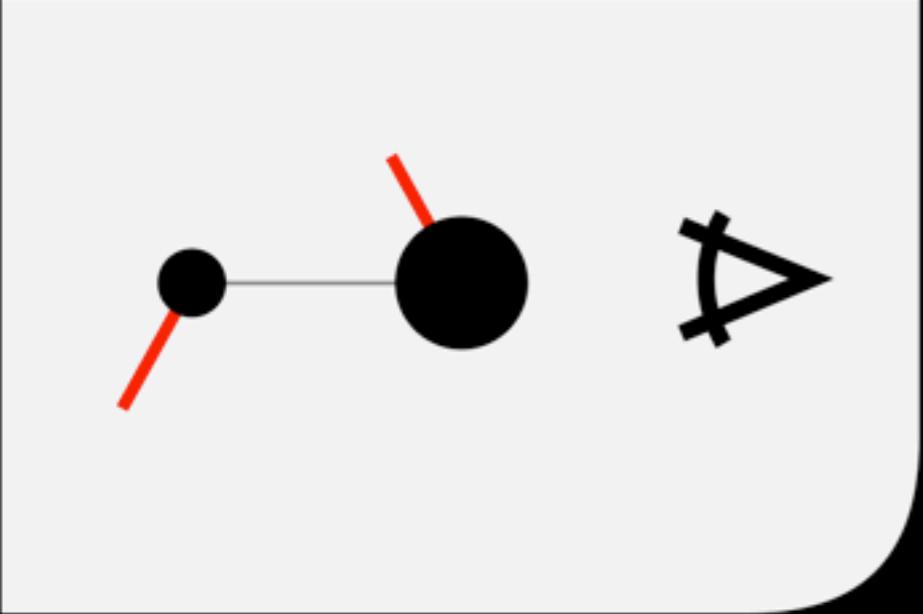
Hannam et al. (2013)



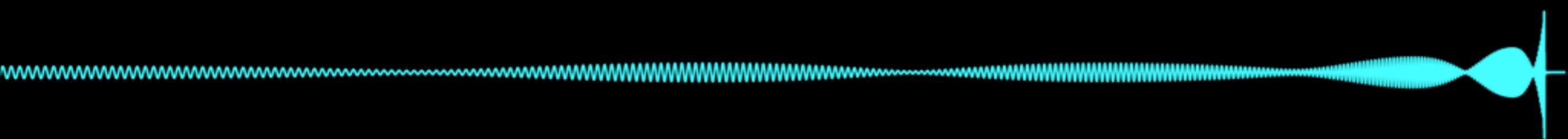
$\nabla$

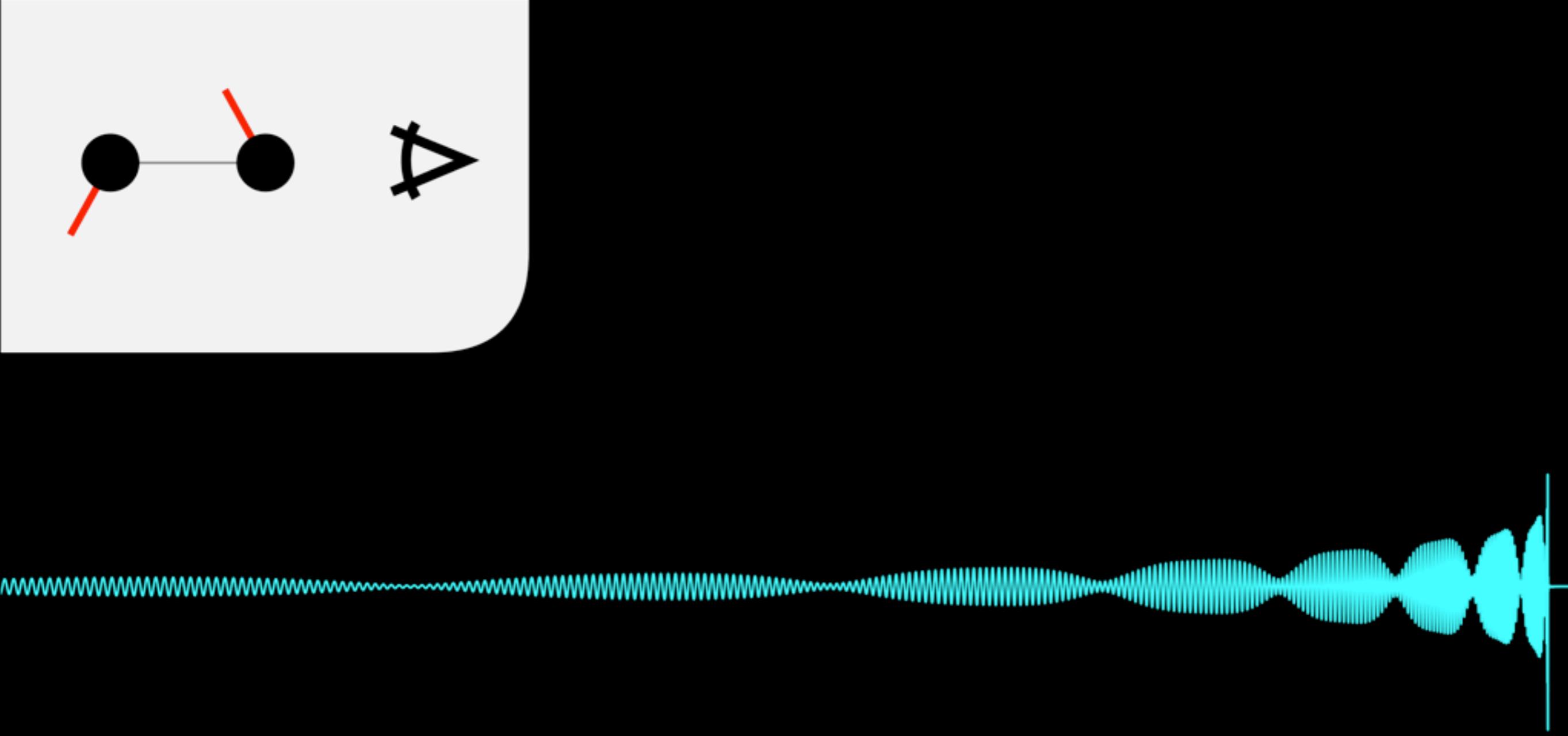






Δ





# OI Models

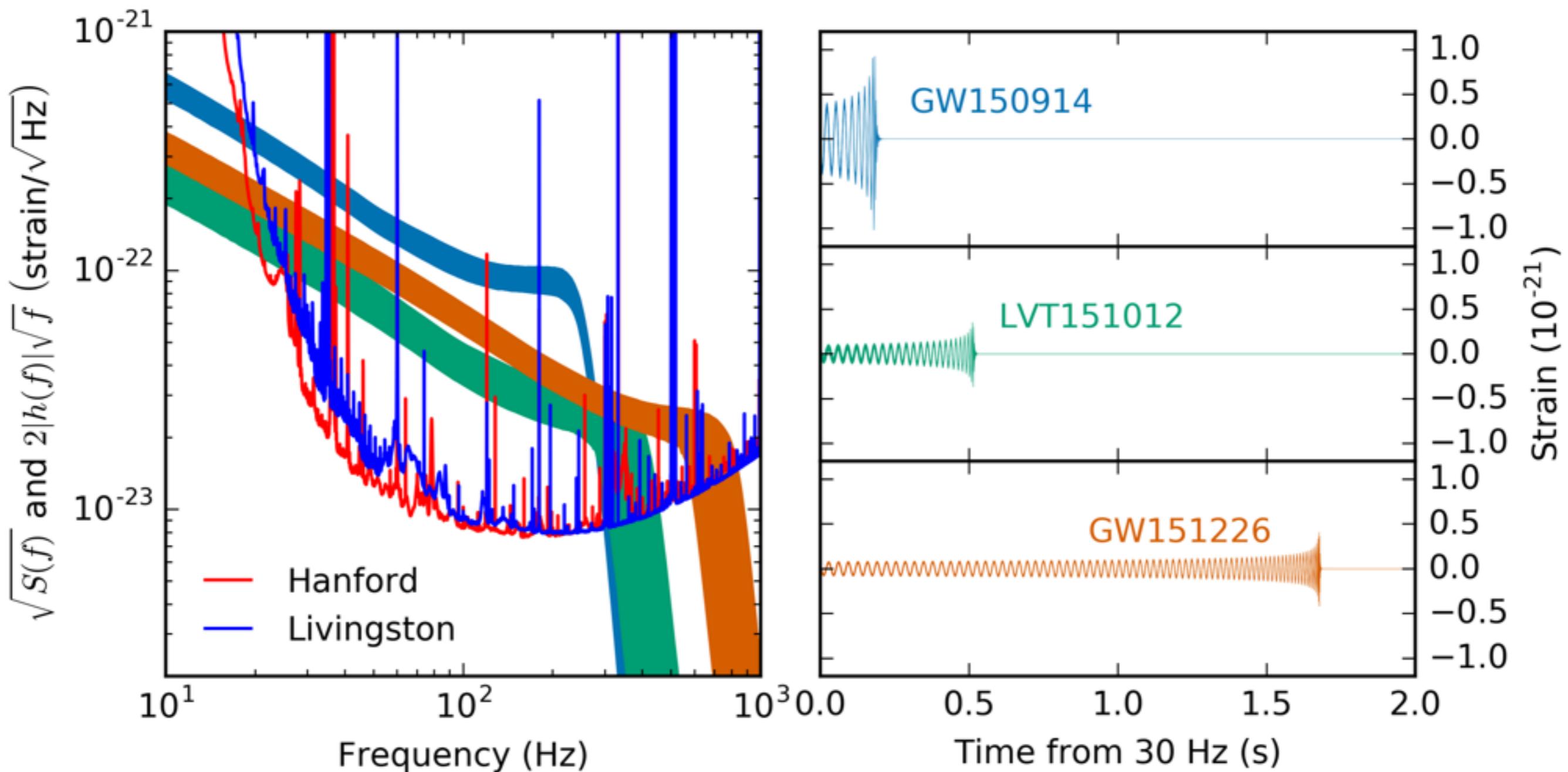
**SEOBNR(v2): double-spin, non-precessing**

**IMRPhenomP(v2): precessing, effective spins**

**SEOBNR(v3): precessing, double-spin**

# BBHs in OI

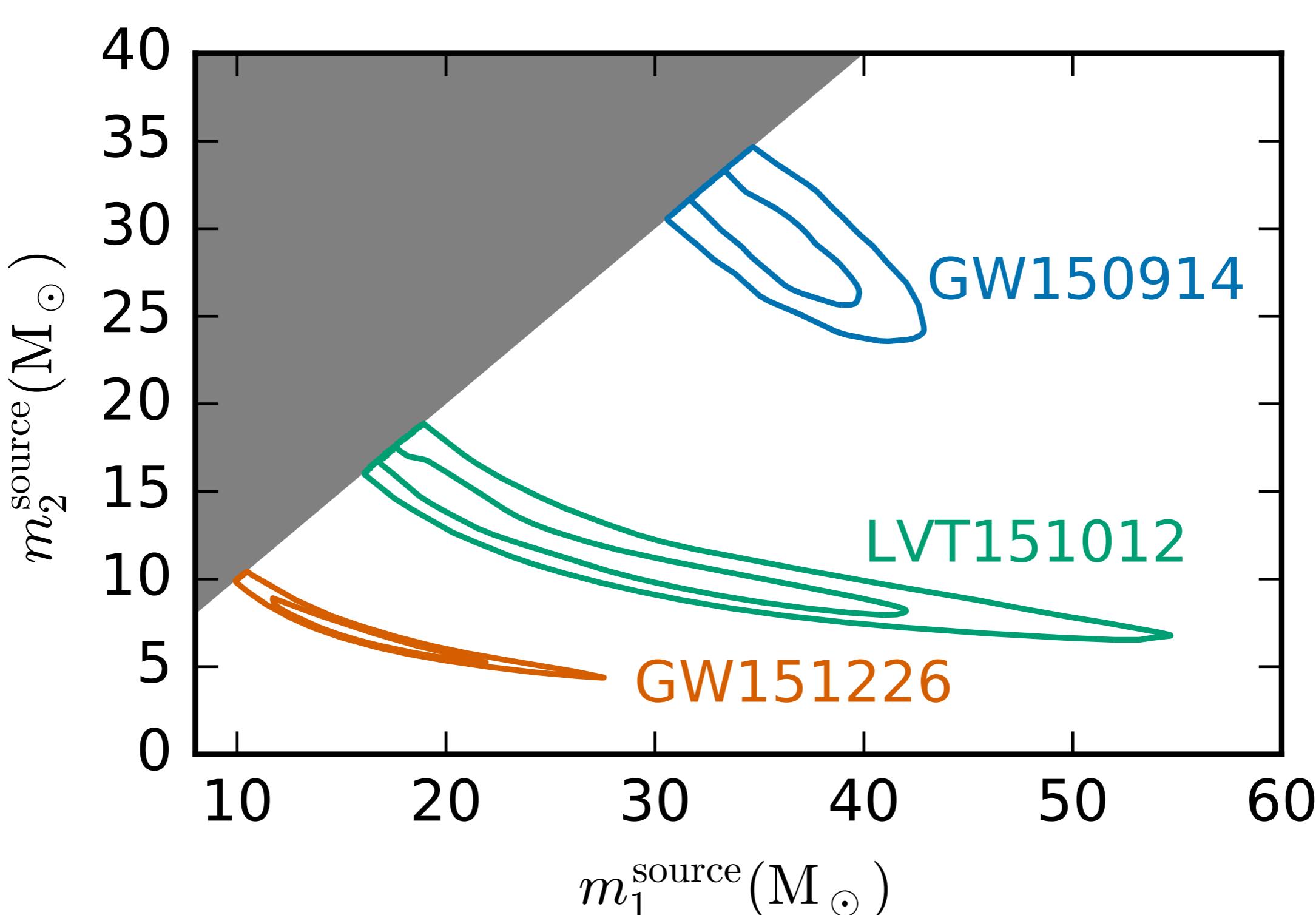
non-precessing



Abbott et al. (2016)

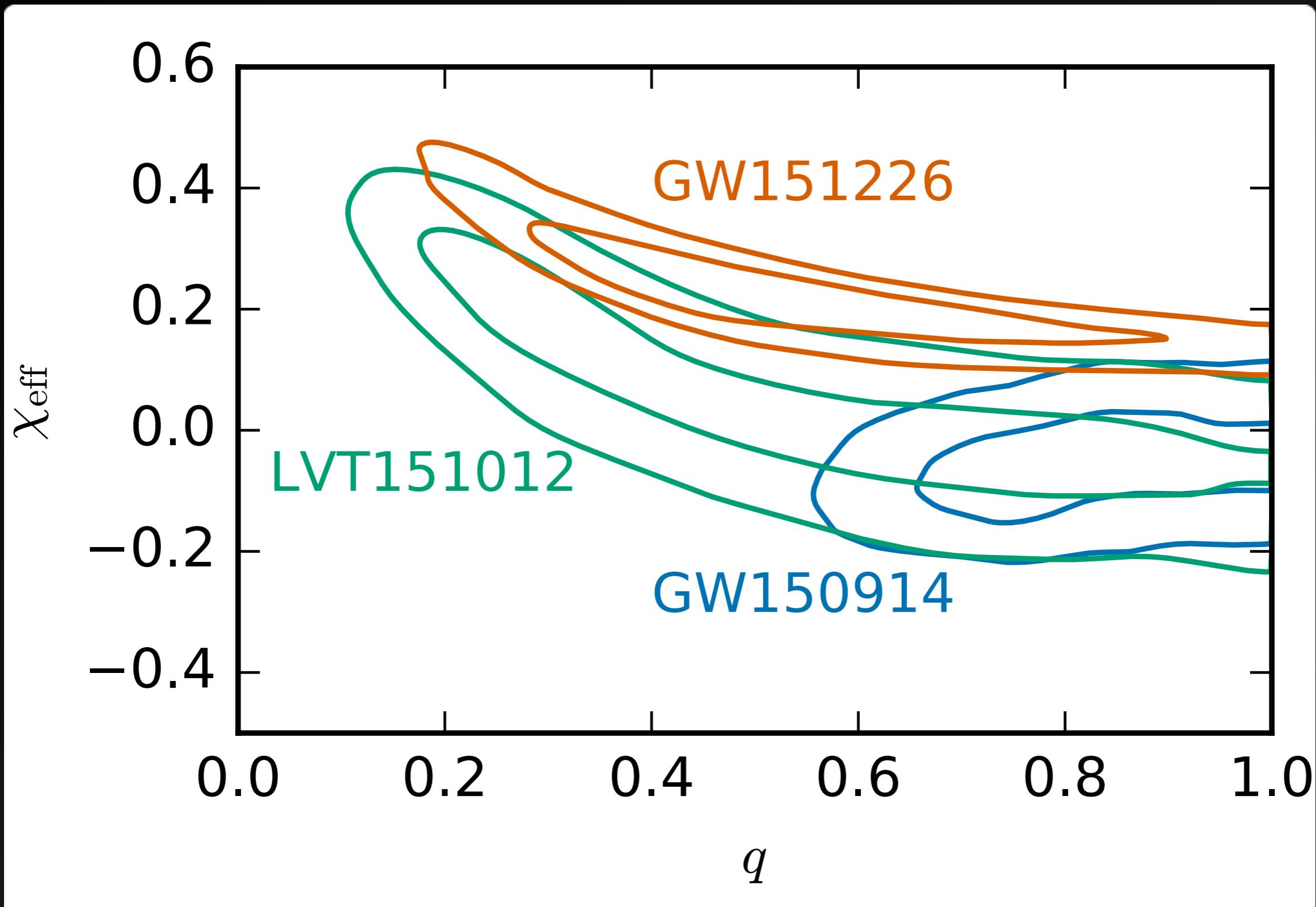
Ben Farr

# Black Hole Masses



Abbott et al. (2016):  
PRX 6, 041015

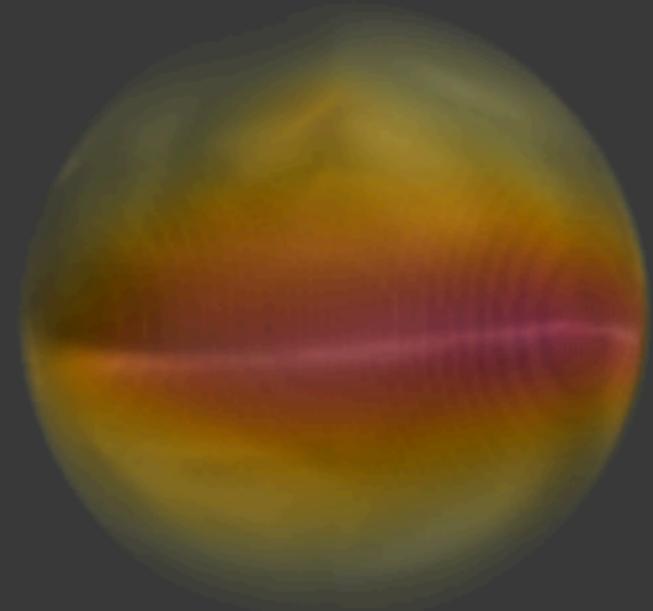
# Mass-Spin Degeneracy



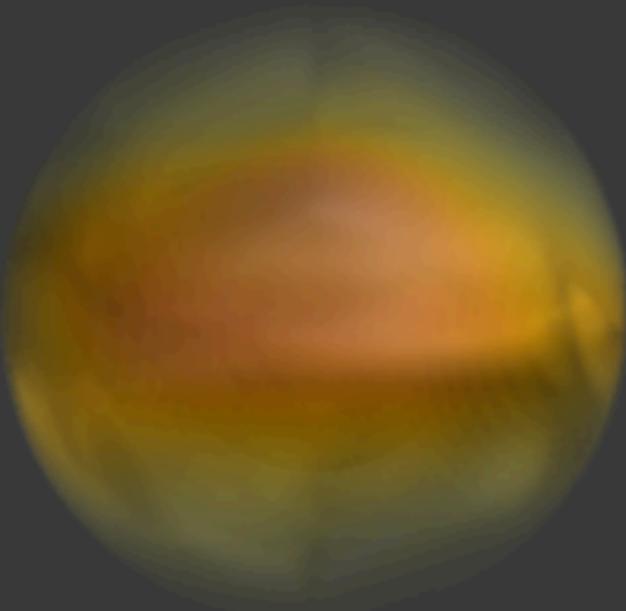
Abbott et al. (2016):  
PRX 6, 041015

# OI Spin Constraints

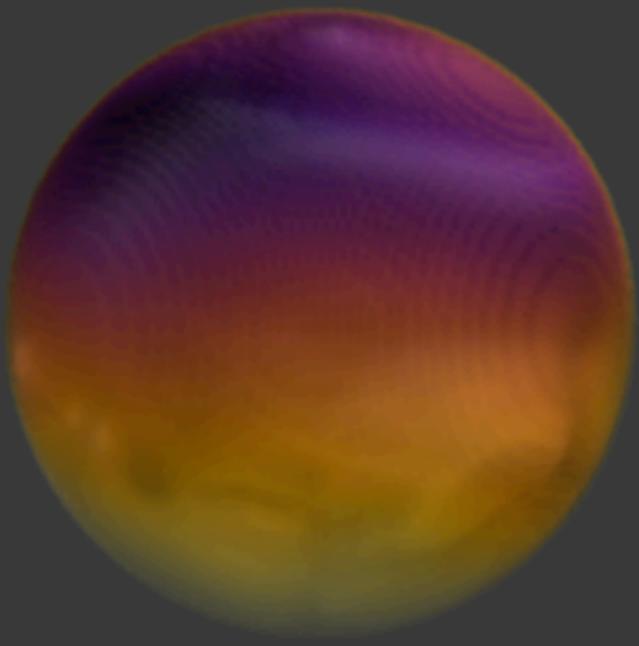
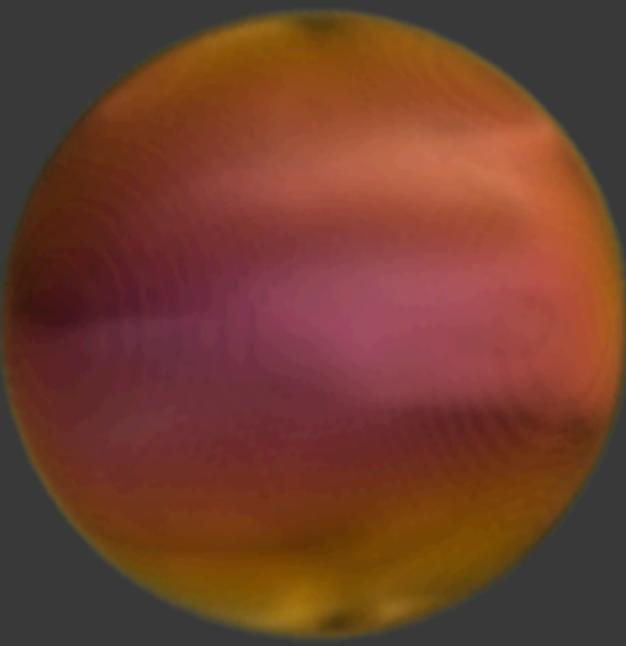
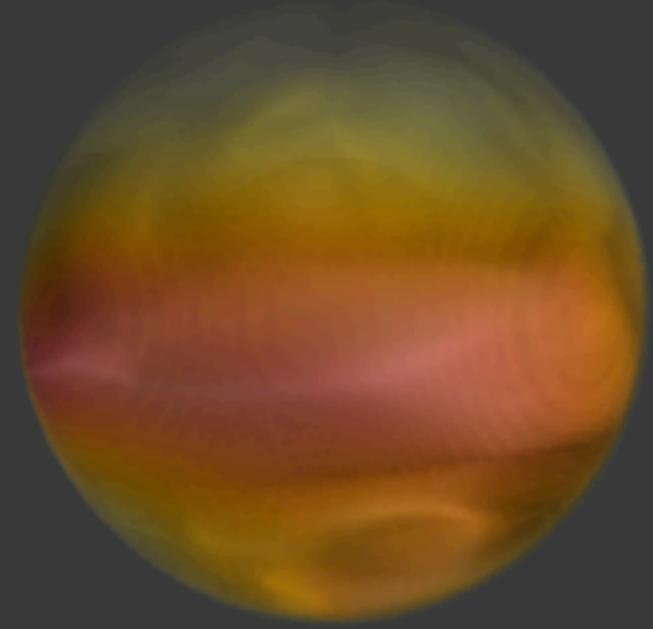
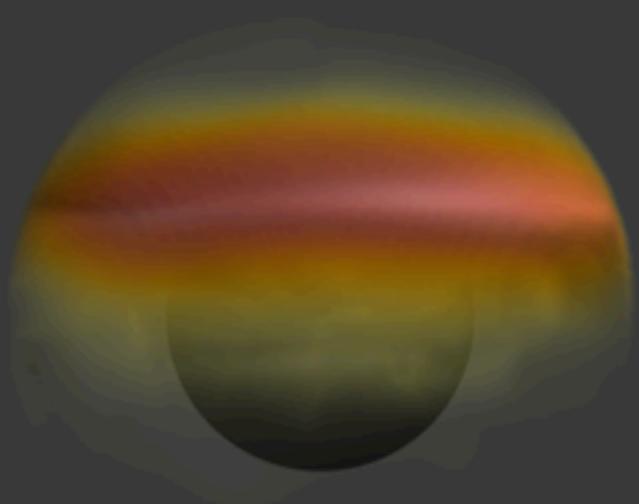
GW150914



LVT151012



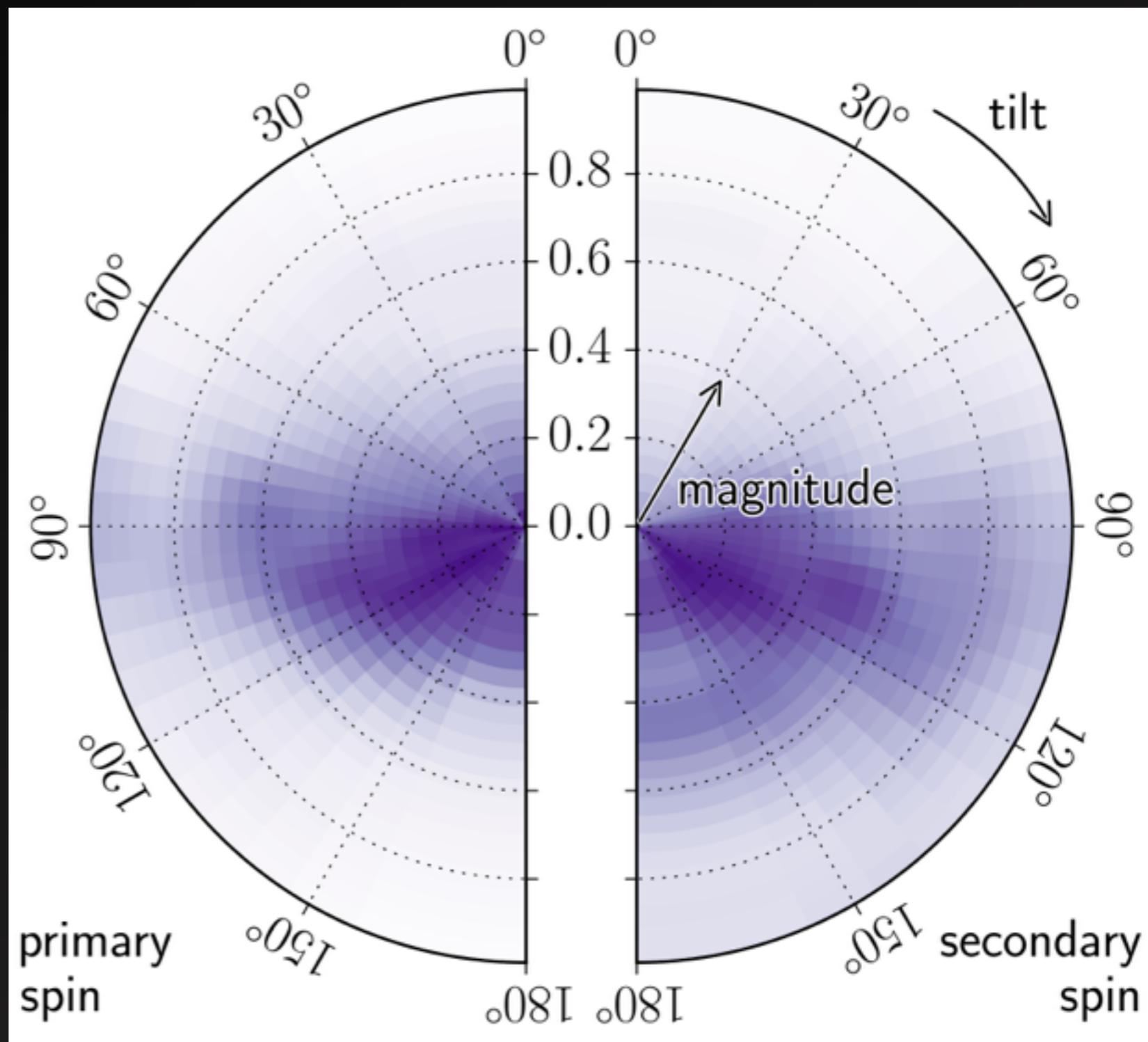
GW151226



# Black Hole Spins

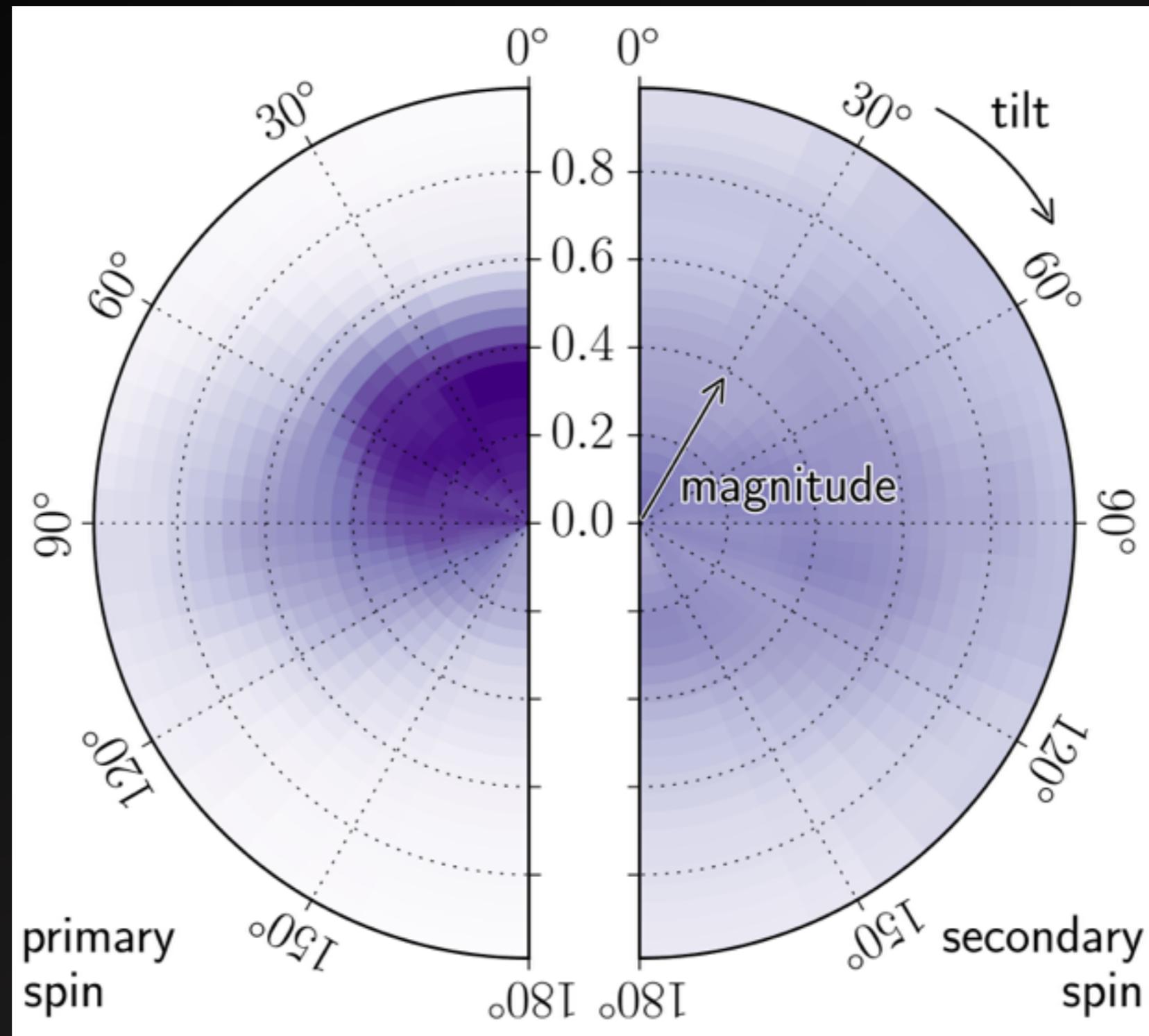
GW150914

BH spin  
not extremal



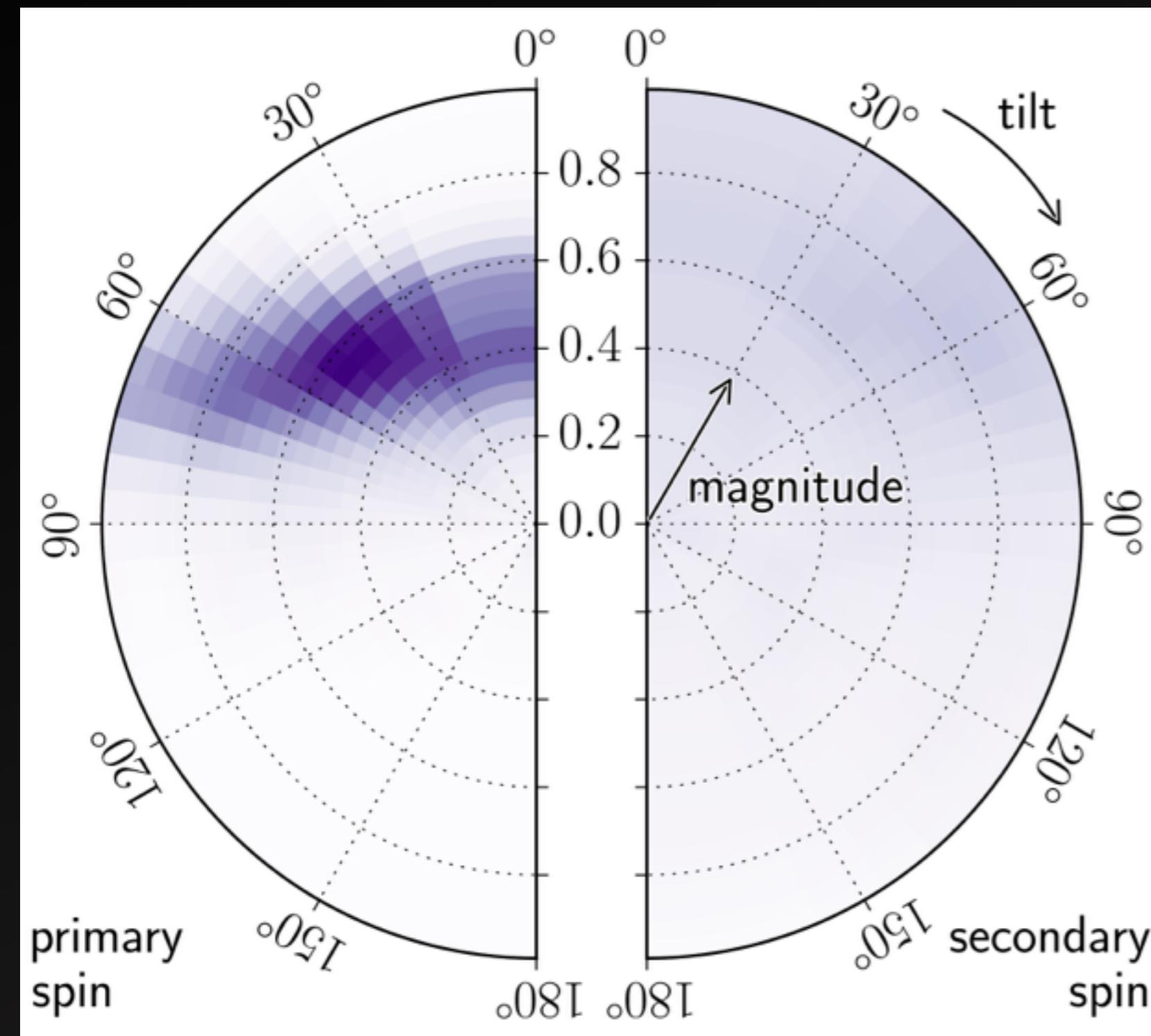
# Black Hole Spins

LVT151012



# Black Hole Spins

GW151226



At least one  
spinning BH

# Next Step: Population Inference

Goal: Use multiple detections to infer population characteristics

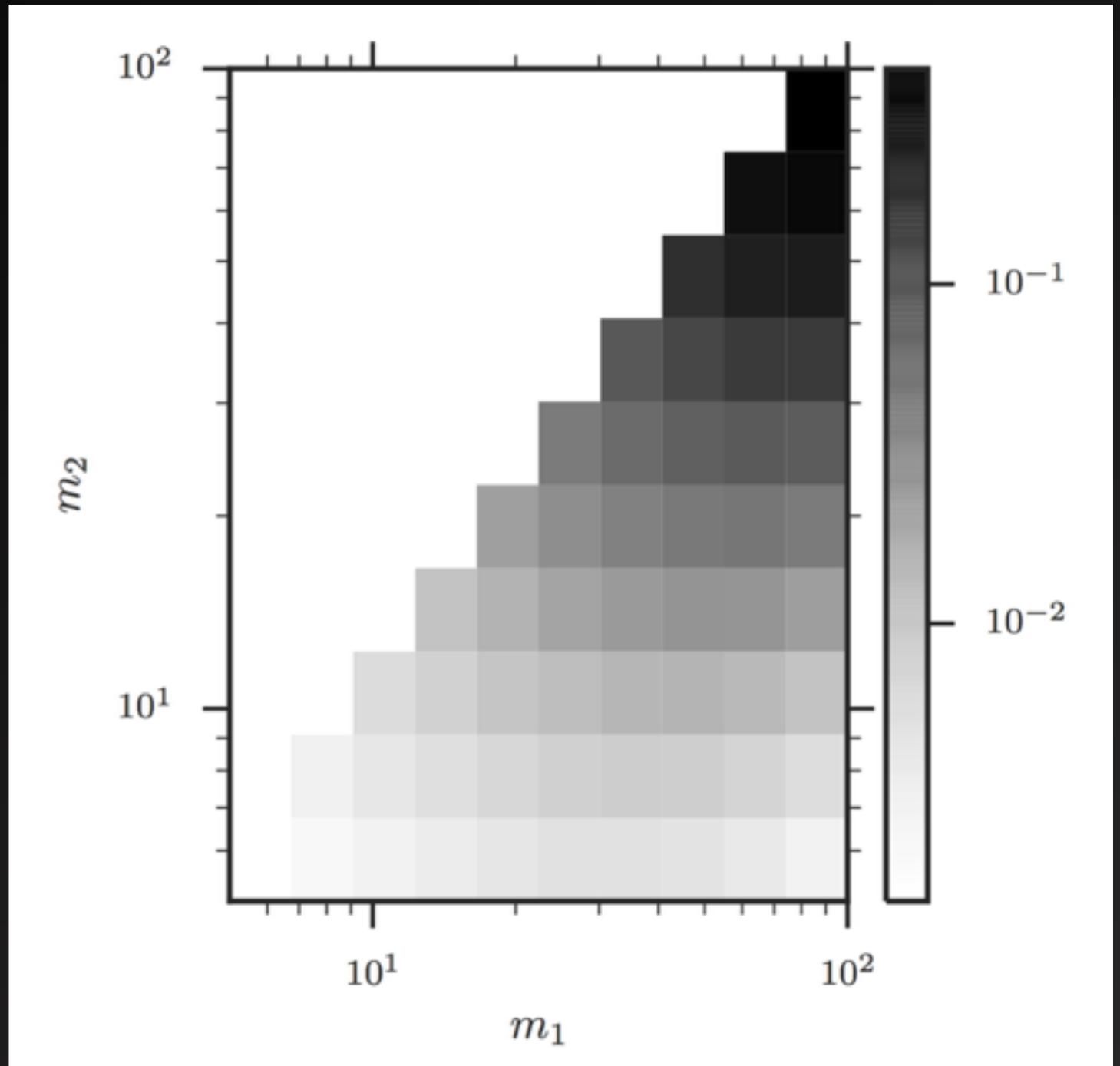
Challenges:

- We never precisely characterize sources
- Detections may sparsely cover parameters space

# Population Inference

Parameterized Model:

$$\Gamma_{\alpha}(\theta) \begin{cases} \exp(\alpha_1) & \theta \in \Delta_1 \\ \exp(\alpha_2) & \theta \in \Delta_2 \\ \dots \\ \exp(\alpha_K) & \theta \in \Delta_K \\ 0 & \text{otherwise} \end{cases}$$



# Gaussian Process Prior

Gaussian process to regularize the histogram

$$\begin{aligned} p(\alpha) &= p(\alpha | \mu, \lambda) \\ &= \mathcal{N}[\alpha; \mu, K(\{\Delta_j\}, \lambda)] \end{aligned}$$

$\mu, \lambda :$  hyperparameters for the mean and length scales of the GP,

# Hierarchical Bayesian Inference

Population posterior density:

$$p(\vec{\alpha}|\{d_i\}) \propto p(\vec{\alpha}) \int p(\{d_i\}|\{\theta_i\}) p(\{\theta_i\}|\vec{\alpha}) d\{\theta_i\}$$

Gaussian process prior

Single-event likelihood

Single-event (population-based) prior

$$p(\{\theta_i\}|\alpha) = \exp \left( - \int \hat{\Gamma}_\alpha(\theta) d\theta \right) \prod_i \hat{\Gamma}_\alpha(\theta_i)$$

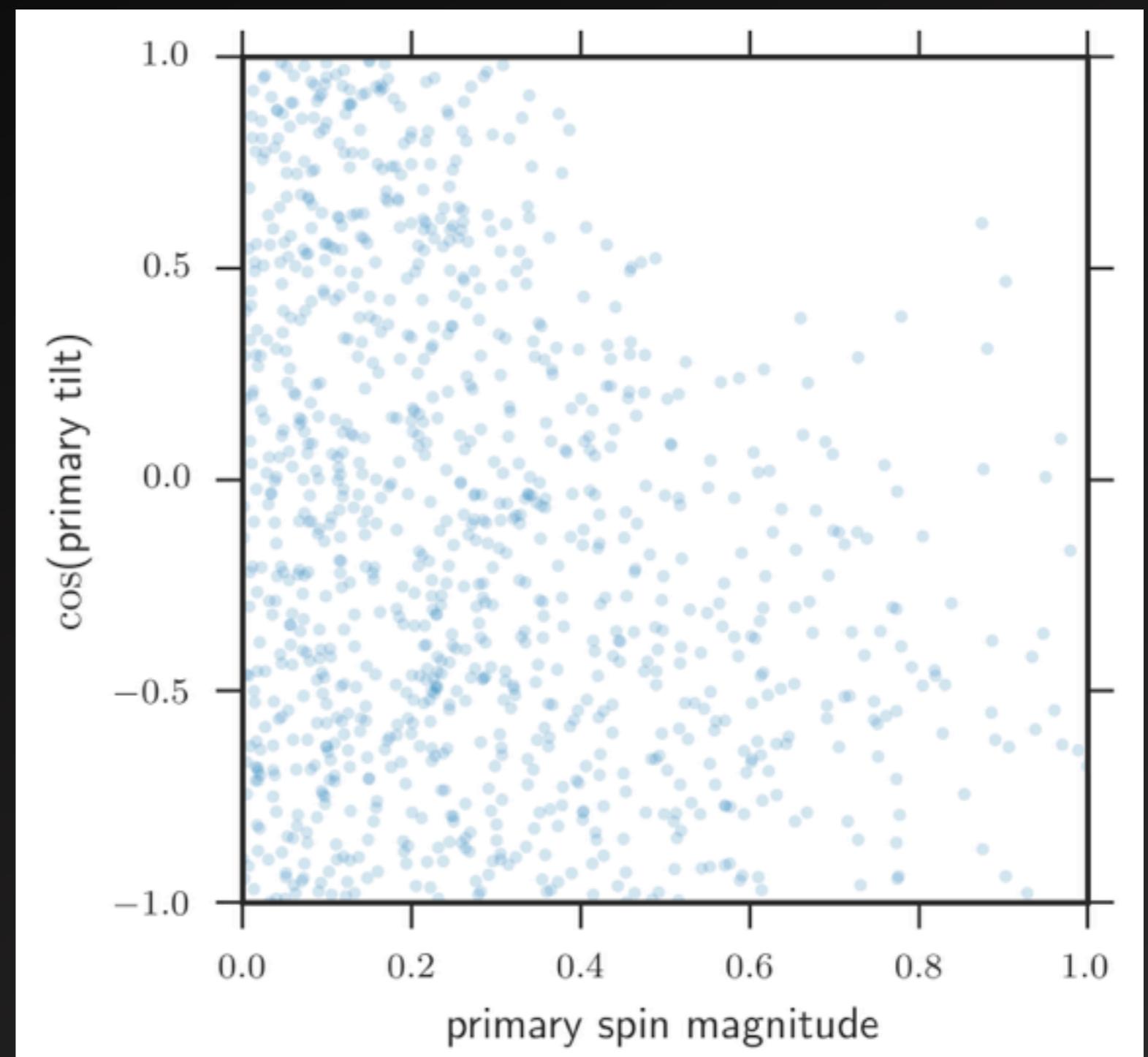
Hogg et al. (2010)

Mandel, Farr, Colonna et al. (2016)

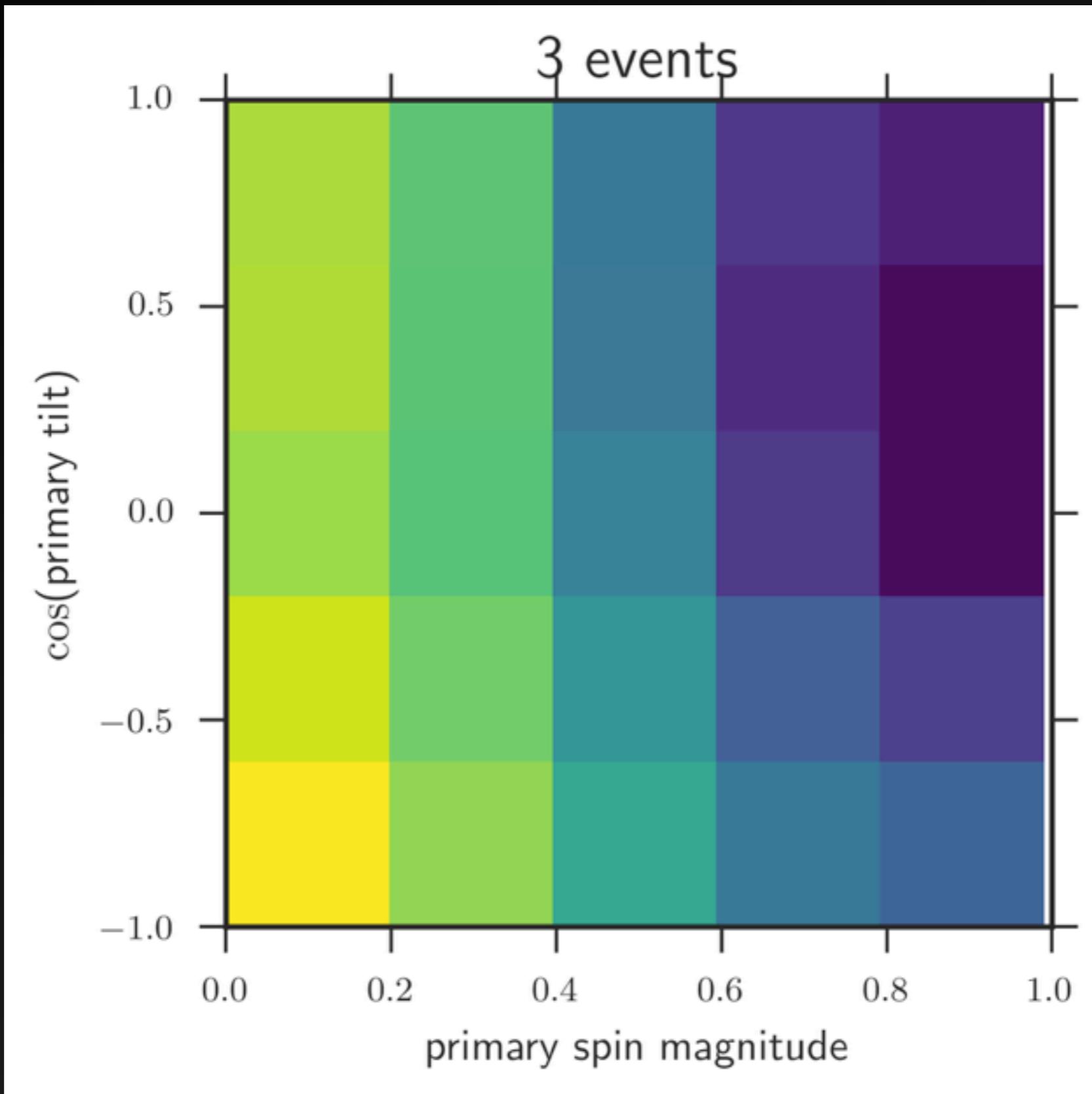
# Spin Distribution

Simulated distribution:  
primary spin  $< 0.5$   
primary tilt  $< 45^\circ$

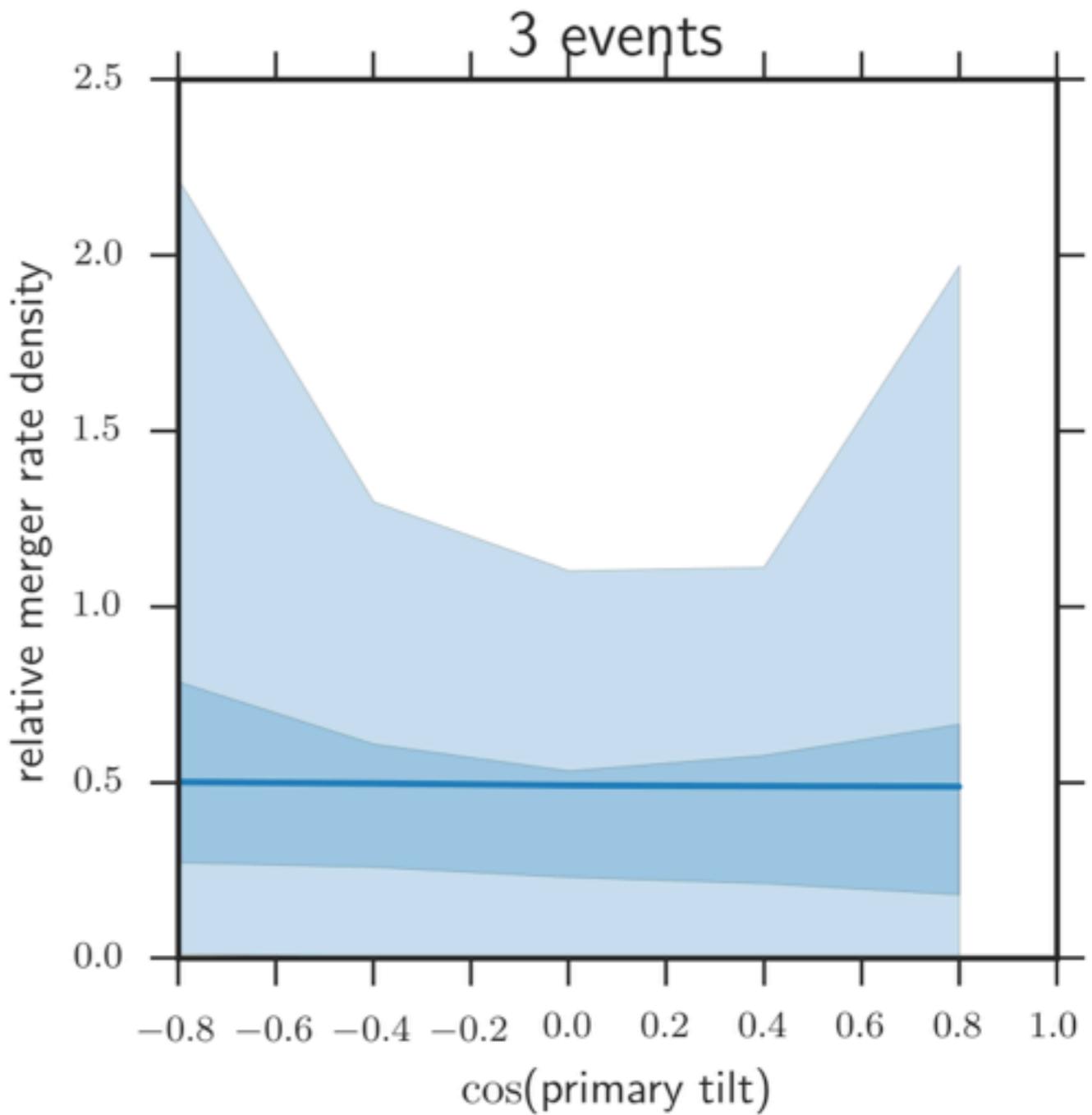
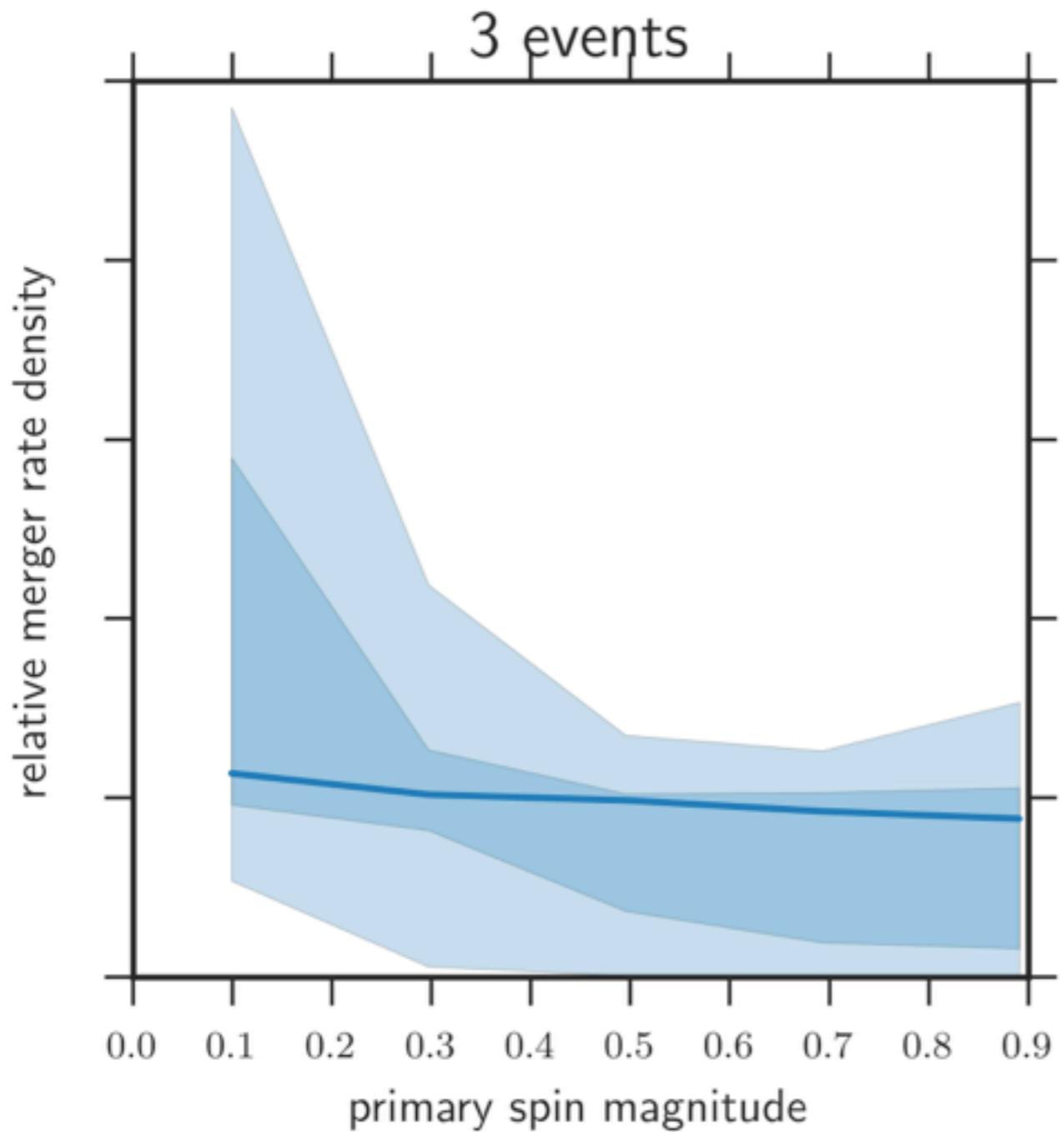
Individual Event Posteriors:



# Spin Distribution



# Spin Distribution



# Spin Distribution

Support for non-uniform distribution of spin magnitudes after  $\sim 10$  events.

Rule out isotropic spin distribution after  $\sim 20$  events.

# Conclusions

Interesting population characteristics may be found after only  $\sim 10$  events.

After 10's of events, strong features in the population begin to be constrained.

Possibly by the end of O2, likely by the end of O3, things should get interesting.

# Questions?