



Constraints on the gravitational wave background with pulsar timing arrays

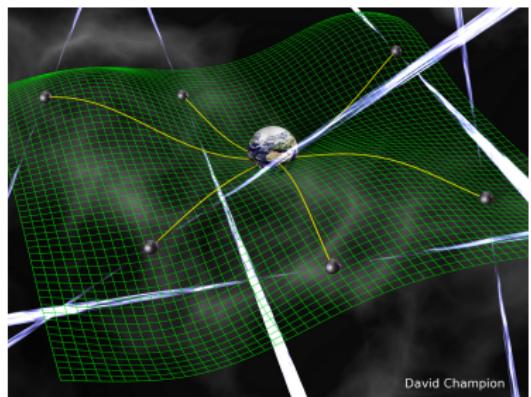
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Cosmic Mergers Workshop – 22 September 2017

Overview

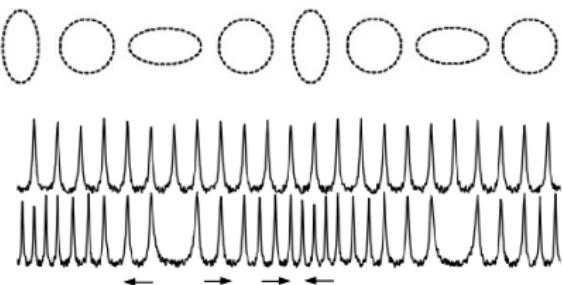
- Pulsar timing searches for gravitational wave background
- No detection yet, but upper limits reaching astrophysically interesting sensitivities
- Recent doubt cast on binary assembly theories
→ are mergers stalling / accelerated??
- Bayesian analysis with astrophysical prior
- Are prediction consistent with upper limits?
→ yes so far!

Pulsar timing arrays

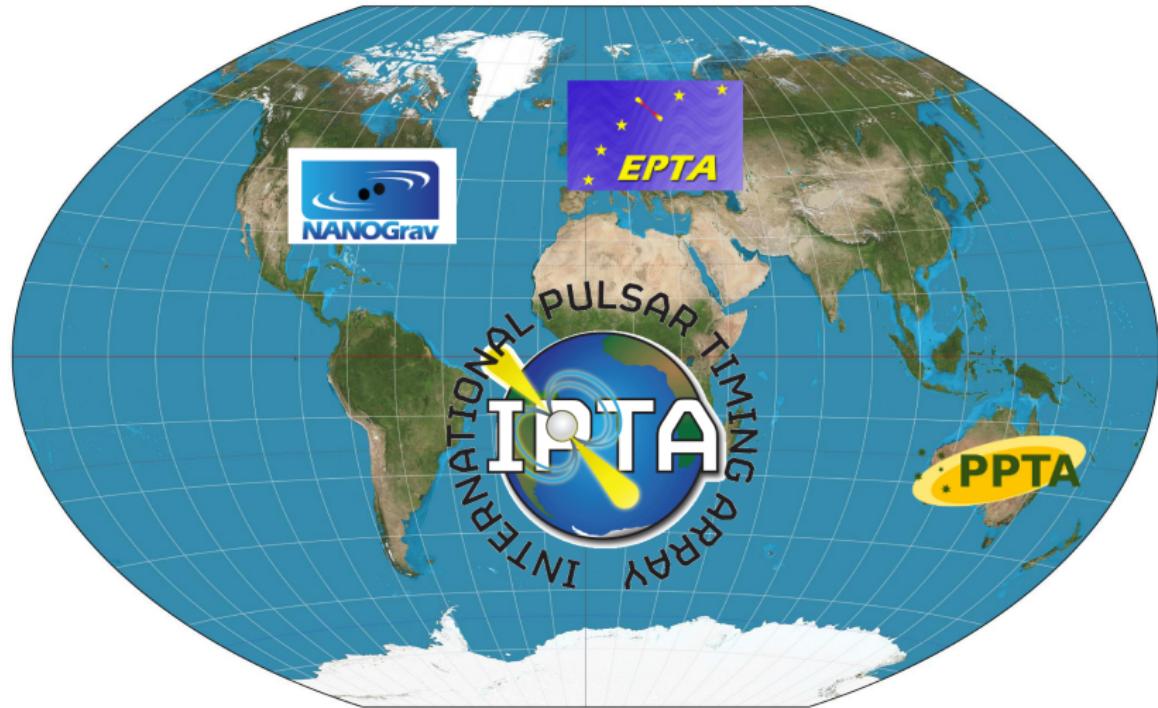


Pulsar timing array

- Gravitational wave background from many mergers
- At nHz frequencies
- Search for deviations in pulse time of arrivals



Searching for the background



Recent results – data

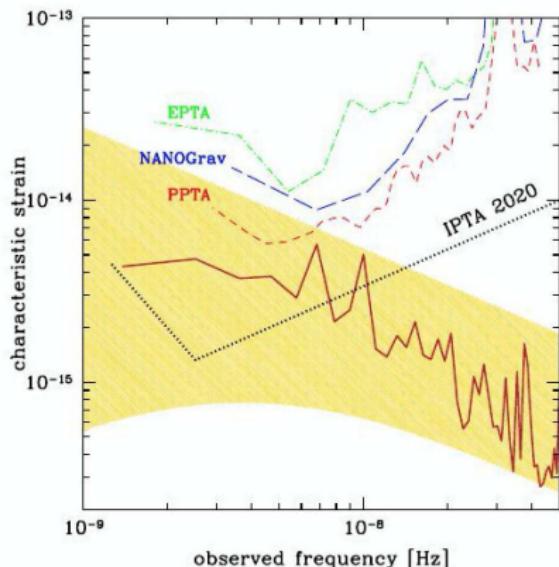


Image: A. Sesana
(reproduced from Hobbs & Dai 2017)

- Most stringent upper limit from Parkes Pulsar Timing array (Shannon et al 2015)

- $h_{ul} < 1 \times 10^{-15}$ at 95% confidence ($f = 1/1\text{yr}$)

- Are predictions in trouble?
 - Eccentricity?
 - Stalling?

Can we place any constraints on the population?

- Bayesian analysis with astrophysical prior
- Can we make any statements on our current predictions for the gravitational wave background
- Our model:
 - merger rate density
 - chirp mass distribution ($\mathcal{M} = (m_1 m_2)^{3/5} / (m_1 + m_2)^{1/5}$)
 - redshift distribution
 - eccentricity at decoupling from the environment

Model

$$h^2(f) = \frac{4G}{\pi c^2 f} \int_0^\infty dz \int_0^\infty d\mathcal{M} \frac{d^2n}{dzd\mathcal{M}} \frac{dE}{df_r}$$

observed GW frequency

sum sources over redshift & chirp mass

number of sources per redshift and chirp mass interval

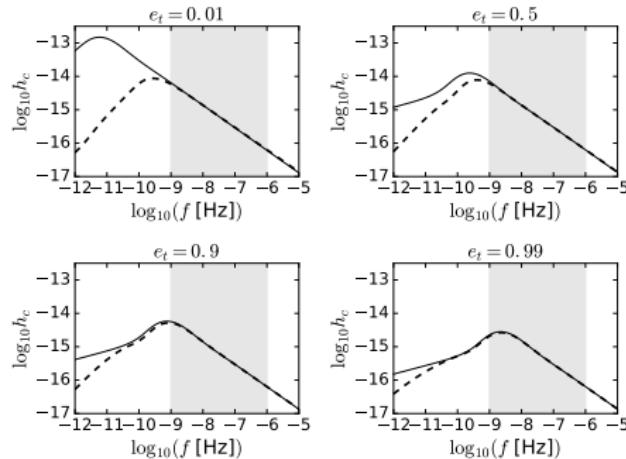
energy spectrum for eccentric binaries

e_t

emitted GW frequency (source frame)

Chen, Sesana & Del Pozzo 2017 (10.1093/mnras/stx1093)

Eccentricity



- Some eccentricity at decoupling
- Population of eccentric gravitational wave driven binaries
- Environmental influence affects lower than PTA band
- Depletes sources at low frequency
- This is the **same** for all binaries

Chen, Sesana & Del Pozzo 2017 (10.1093/mnras/stx1093)

Model

$$h^2(f) = \frac{4G}{\pi c^2 f} \int_0^\infty dz \int_0^\infty d\mathcal{M} \frac{d^2 n}{dz d\mathcal{M}} \frac{dE}{df_r}$$

Diagram illustrating the components of the equation:

- observed GW frequency**: Points to the term f in the equation.
- sum sources over redshift & chirp mass**: Points to the nested integrals $\int_0^\infty dz \int_0^\infty d\mathcal{M}$.
- number of sources per redshift and chirp mass interval**: Points to the term $\frac{d^2 n}{dz d\mathcal{M}}$.
- energy spectrum for eccentric binaries**: Points to the term $\frac{dE}{df_r}$.
- emitted GW frequency (source frame)**: Points to the variable e_t associated with the frequency df_r .

Model

$$\frac{d^2n}{dzd\mathcal{M}} = \dot{n}_0 \left[\left(\frac{\mathcal{M}}{10^7 M_\odot} \right)^\alpha \exp \left(-\frac{\mathcal{M}}{\mathcal{M}_*} \right) \right] \left[(1+z)^\beta \exp \left(-\frac{z}{z_*} \right) \right] \frac{dt_r}{dz}$$

chirp mass distribution

source frame time

no. mergers ($\text{Mpc}^{-3}\text{Gyr}^{-1}$)

AND e_t

redshift distribution

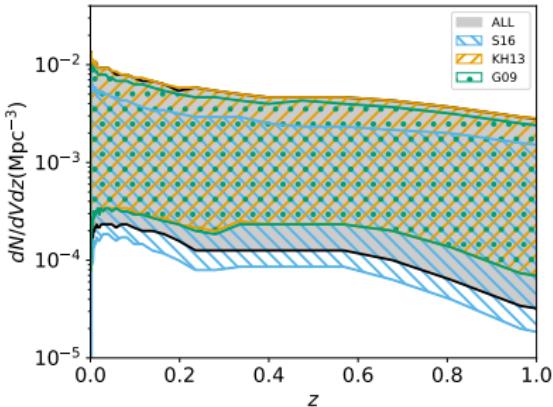
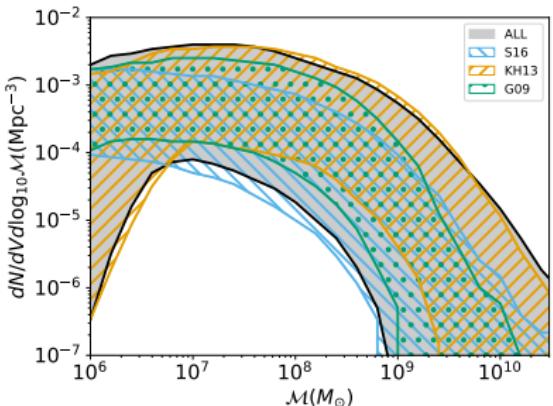
6 parameters model:

- \dot{n}_0 (merger rate density)
- α, \mathcal{M}_* (chirp mass distribution)
- β, z_* (redshift distribution)
- e_t (decoupling eccentricity)

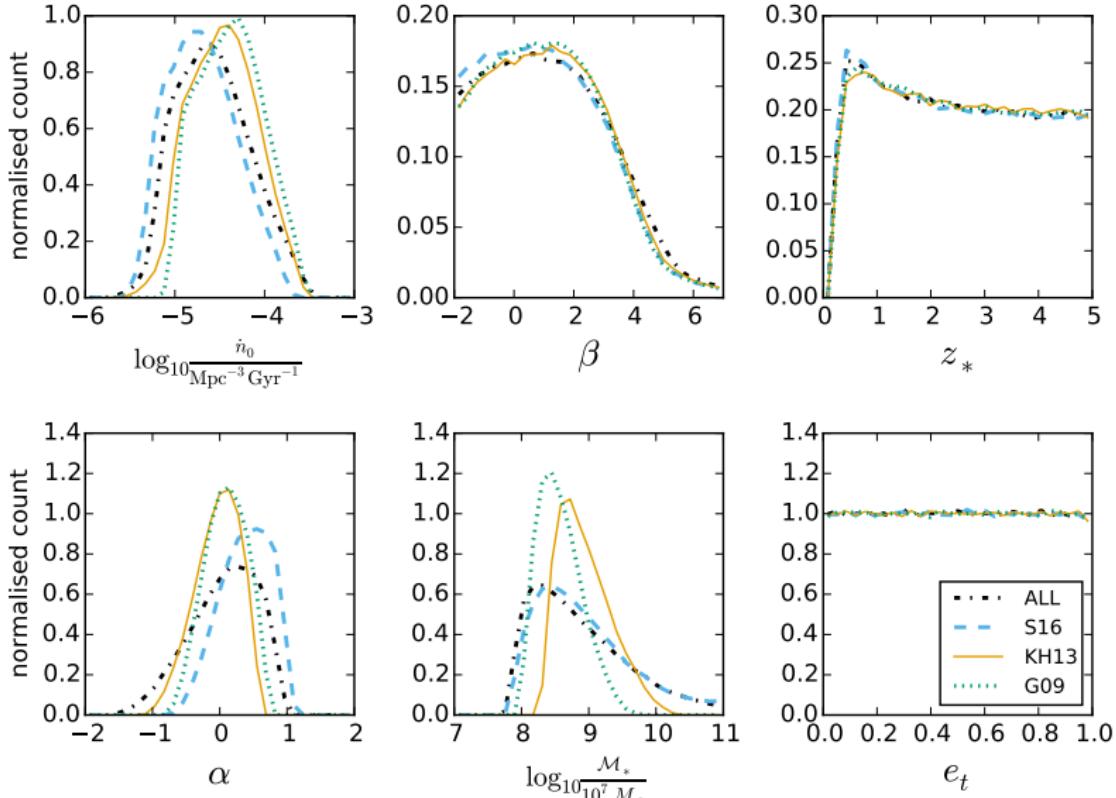
Astrophysical Prior

Model	median strain at $f = 1/1\text{yr}$
Pessimistic	$\approx 4 \times 10^{-16}$
Middling	$\approx 7 \times 10^{-16}$
Optimistic	$\approx 1.5 \times 10^{-15}$
All+ ≈	8×10^{-16}

(Shankar+ 16, Gültekin+ 09,
Kormendy Ho 13)



Astrophysical Prior



Method Overview

$$p(\theta|dM) = \frac{p(\theta|M)p(d|M, \theta)}{p(d|M)}$$

Method Overview

PRIOR
for 6 parameters


$$p(\theta|dM) = \frac{p(\theta|M)p(d|M, \theta)}{p(d|M)}$$

Method Overview

$$p(\theta|dM) = \frac{p(\theta|M)p(d|M, \theta)}{p(d|M)}$$

PRIOR
for 6 parameters

LIKELIHOOD:
upper limit or
detection?

```
graph TD; A[PRIOR for 6 parameters] --> B[p(theta|M)]; C[LIKELIHOOD: upper limit or detection?] --> D[p(d|M, theta)]; E[p(d|M)]
```

Method Overview

$$p(\theta|dM) = \frac{p(\theta|M)p(d|M, \theta)}{p(d|M)}$$

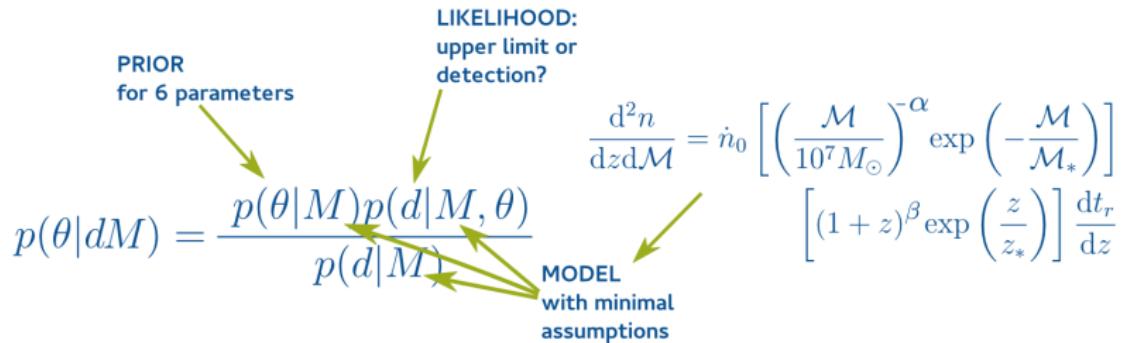
PRIOR
for 6 parameters

LIKELIHOOD:
upper limit or
detection?

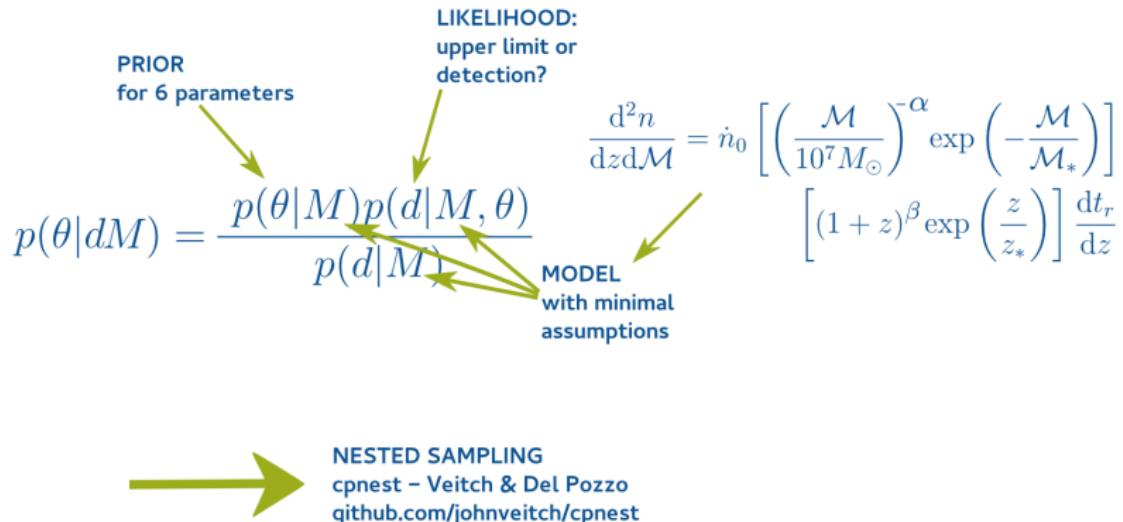
MODEL
with minimal
assumptions

```
graph TD; A[PRIOR for 6 parameters] --> B[p(θ|M)]; C[LIKELIHOOD: upper limit or detection?] --> D[p(d|M, θ)]; E[MODEL with minimal assumptions] --> F[p(d|M)]; G[p(θ|M)p(d|M, θ)/p(d|M)] --> H[ ];
```

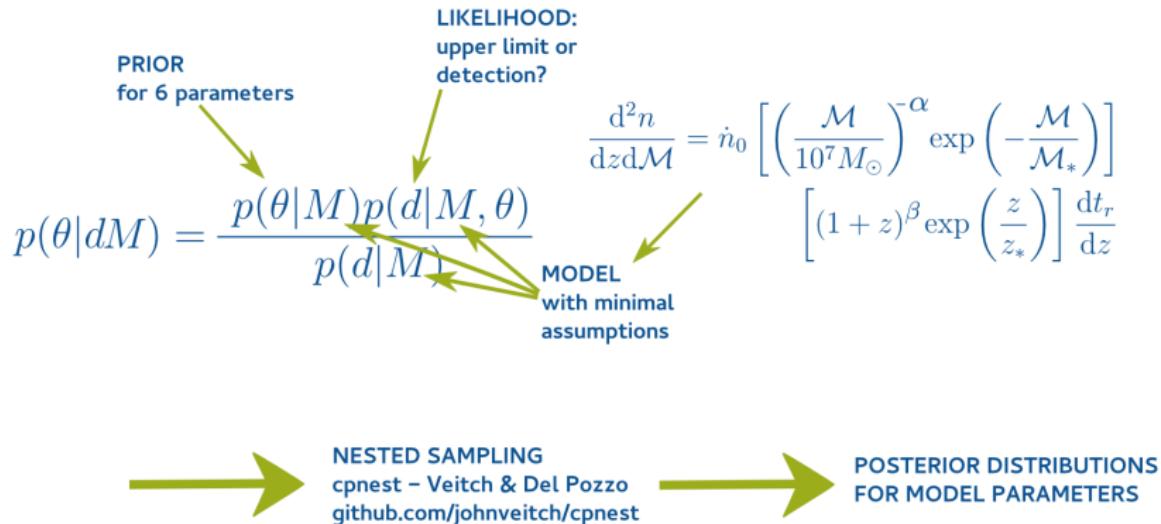
Method Overview



Method Overview

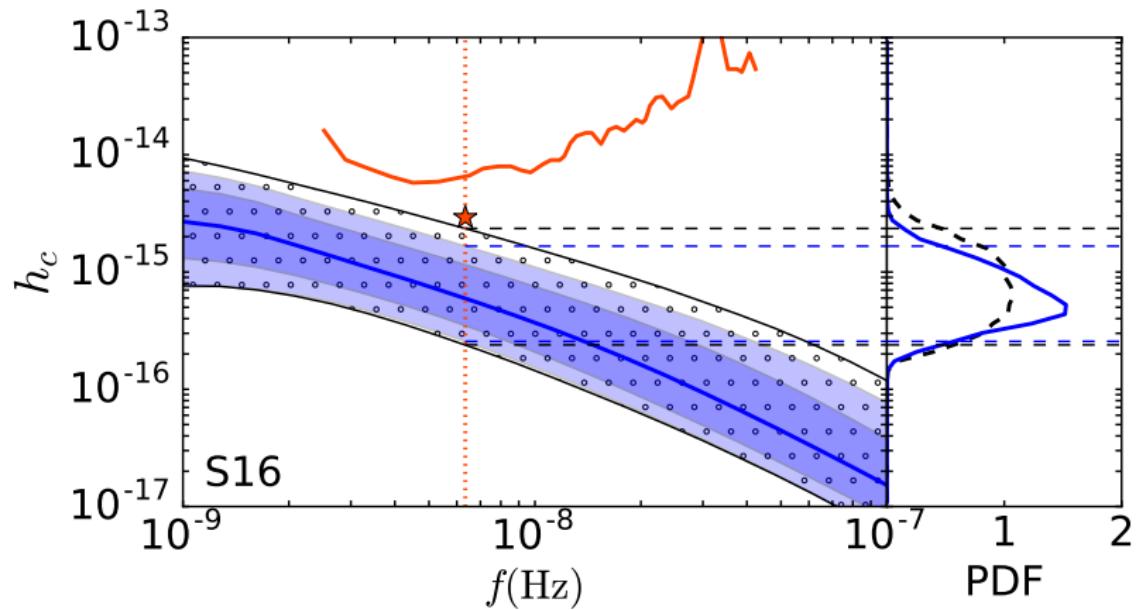


Method Overview



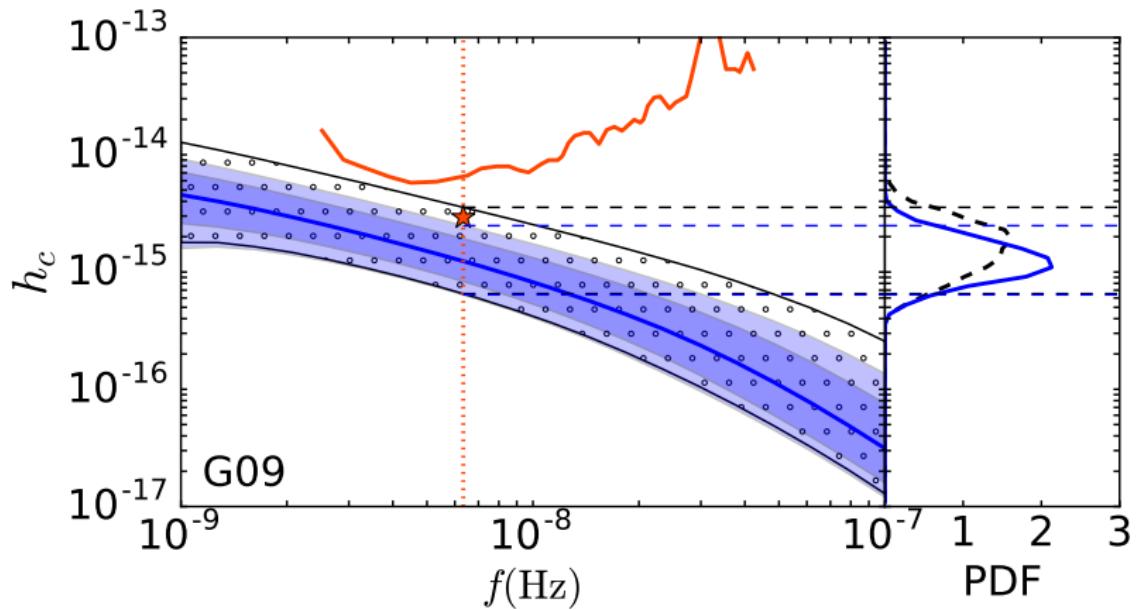
Results

pessimistic



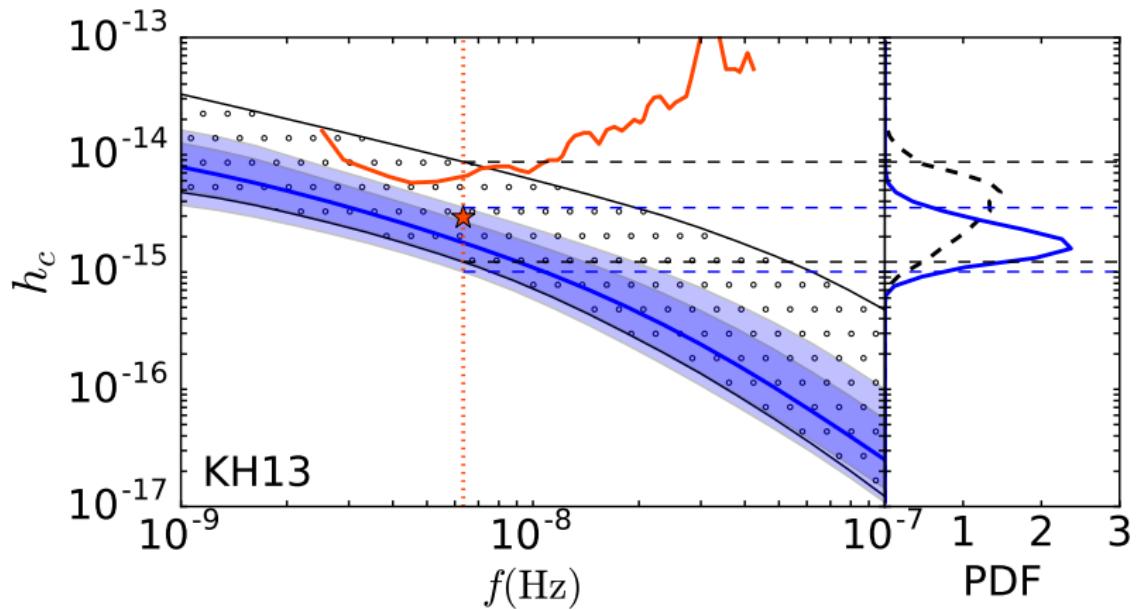
Results

middle



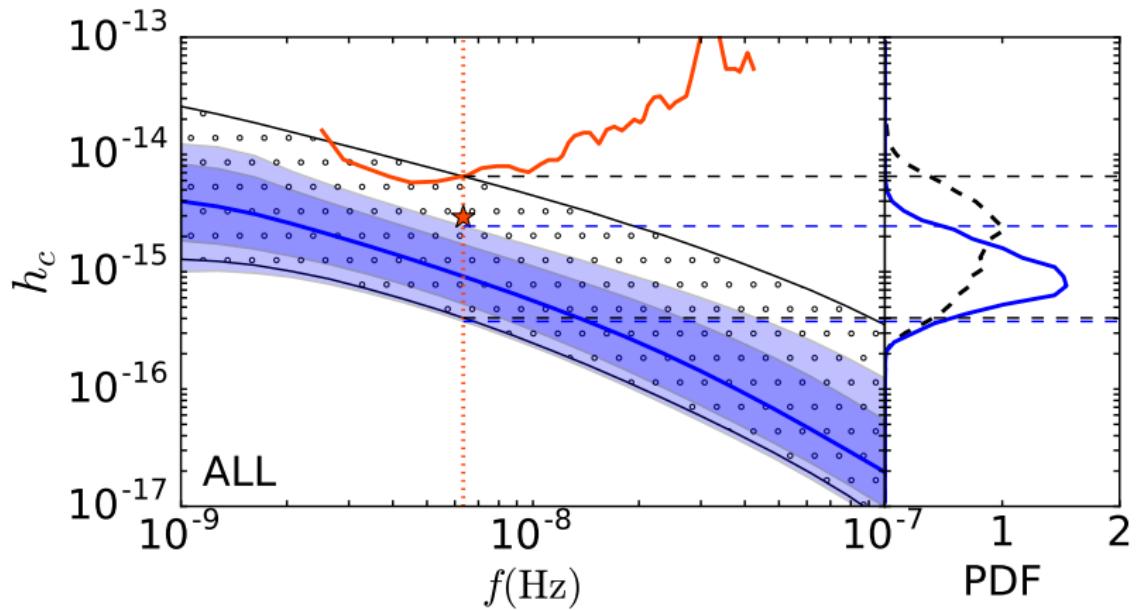
Results

optimistic



Results

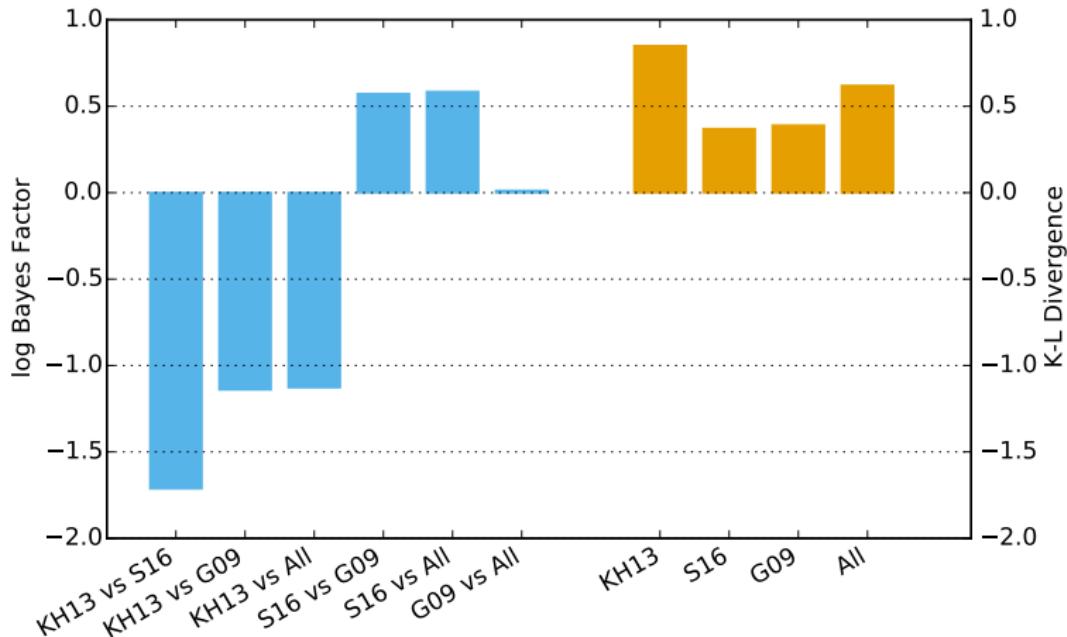
all+



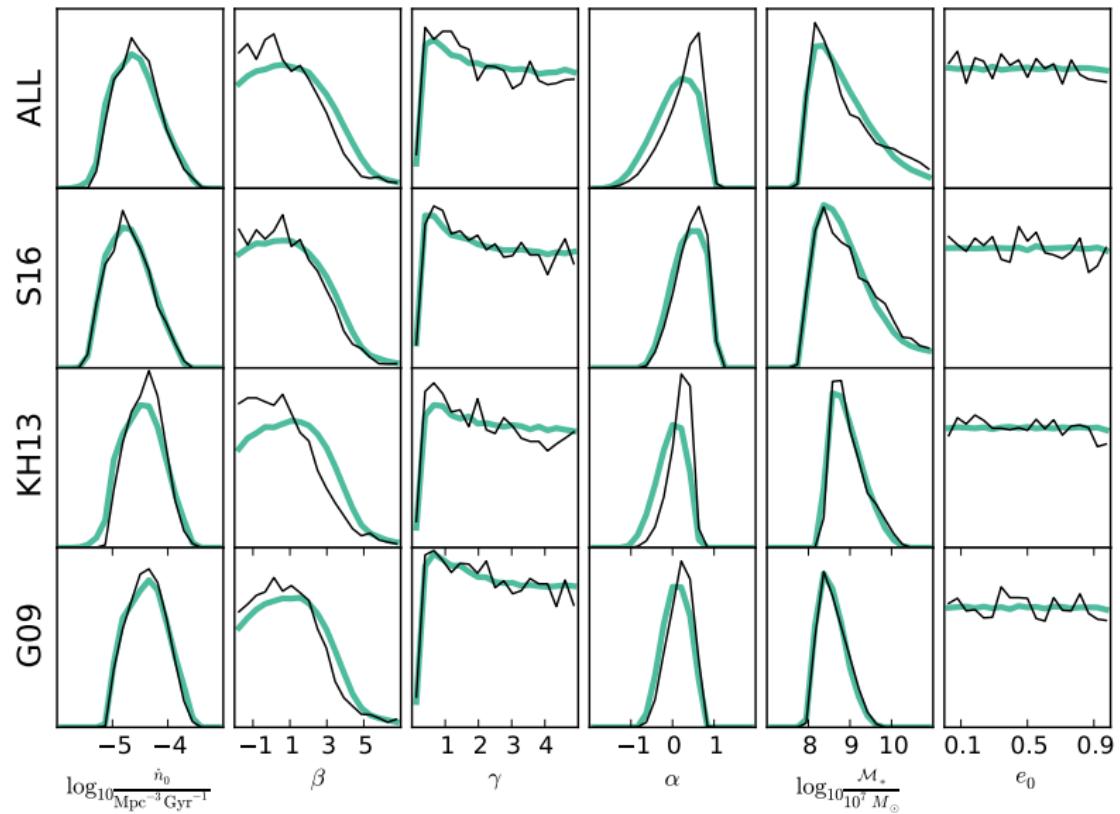
Summary

- Upper limits are eating into predicted values
- But **models are still consistent with observation**
- Little constraint on the model parameters – eccentricity not essential
- No need to rethink predictions yet
- Order of magnitude improvement in sensitivity would put optimistic predictions in more trouble

Bonus slides! Quantitative Results

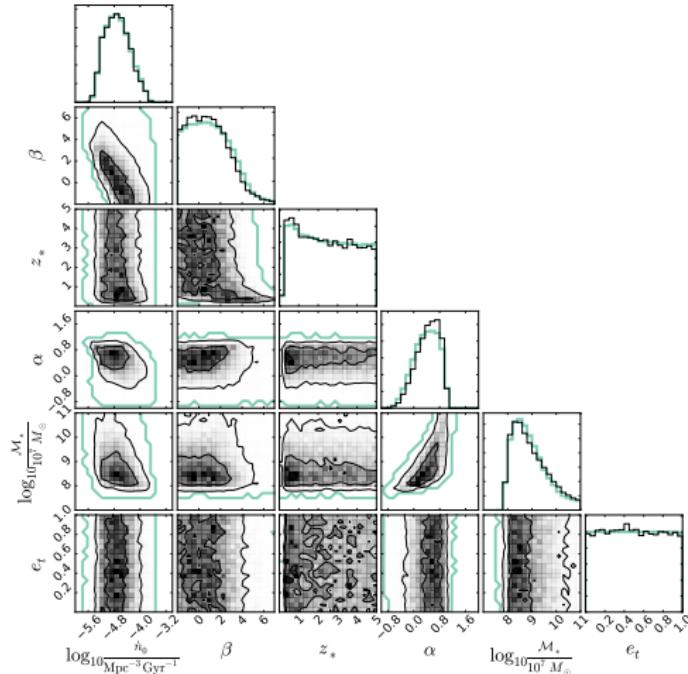


Bonus slides! Parameters



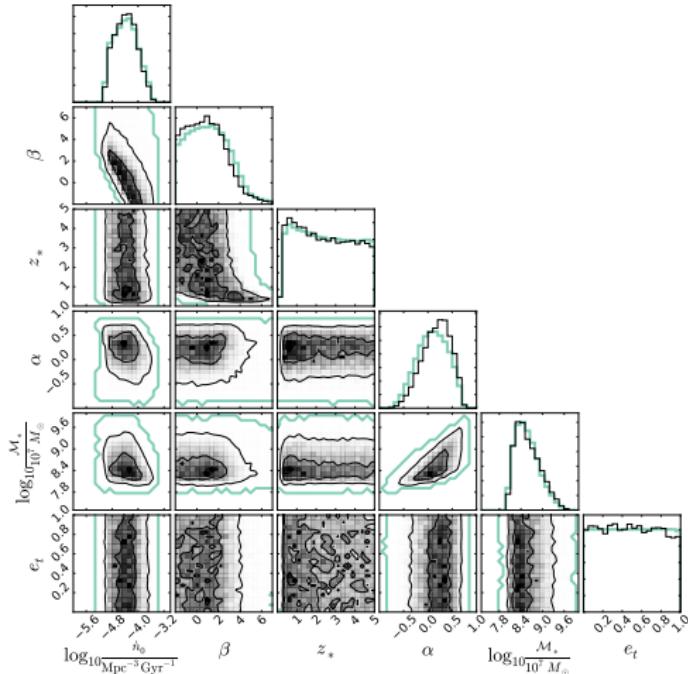
Bonus slides! Parameters

pessimistic



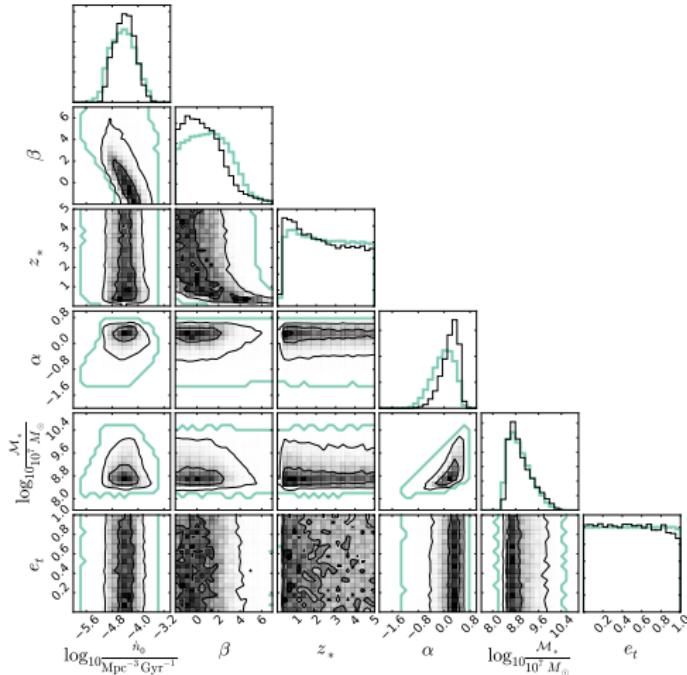
Bonus slides! Parameters

middling



Bonus slides! Parameters

optimistic



Bonus slides! Parameters

all+

