

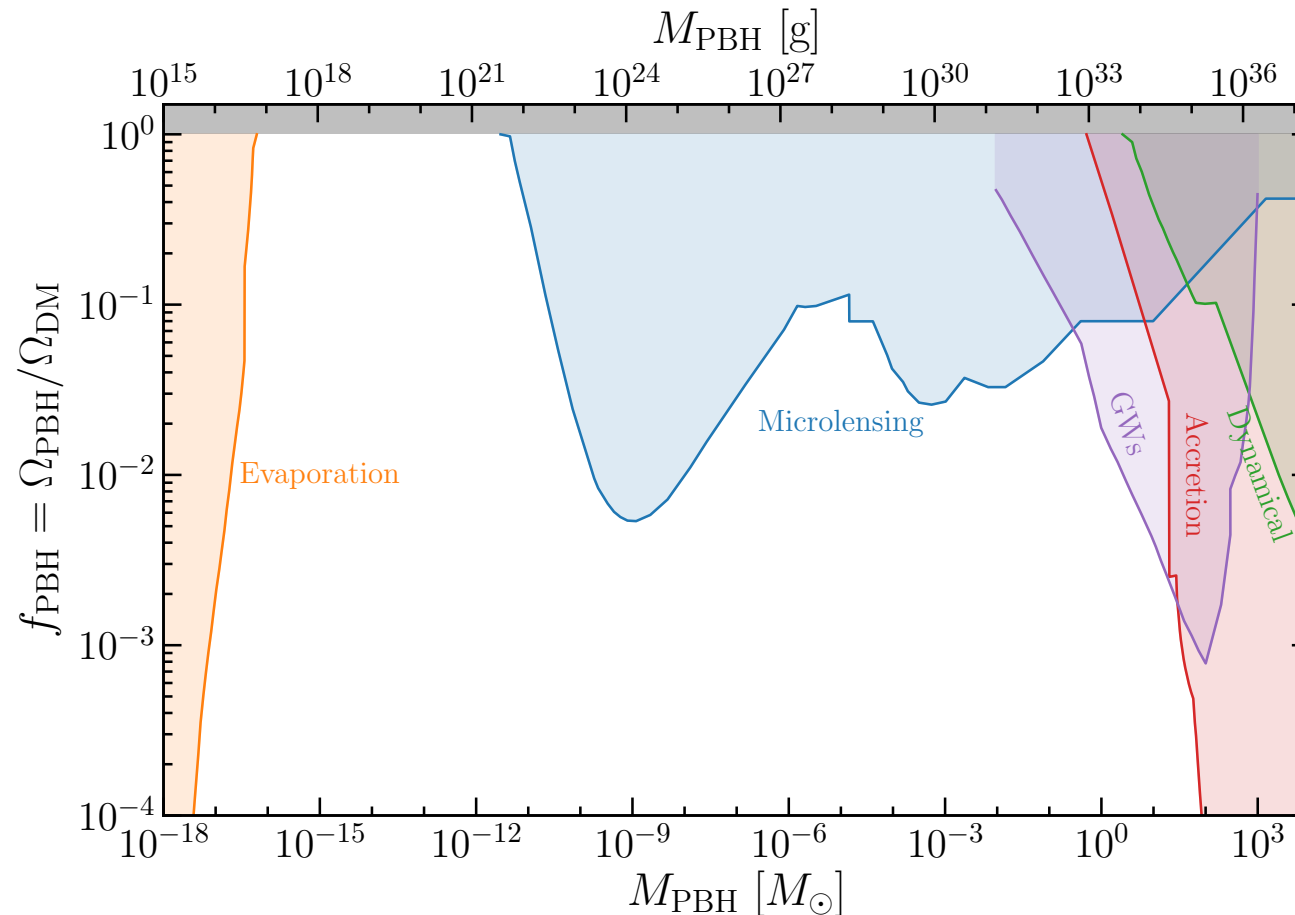
Multi-messenger avenues towards Primordial Black Hole detection



Primordial black holes (PBHs)

Zeldovich 67, Hawking MNRAS 152 (1971), Carr and Hawking MNRAS 168 (1974)

- ▶ BHs can form in early universe from *collapse of overdense regions* $\left(\frac{\delta\rho}{\rho} \gtrsim \frac{1}{3}\right)$
- ▶ Mass related to horizon size at time of formation

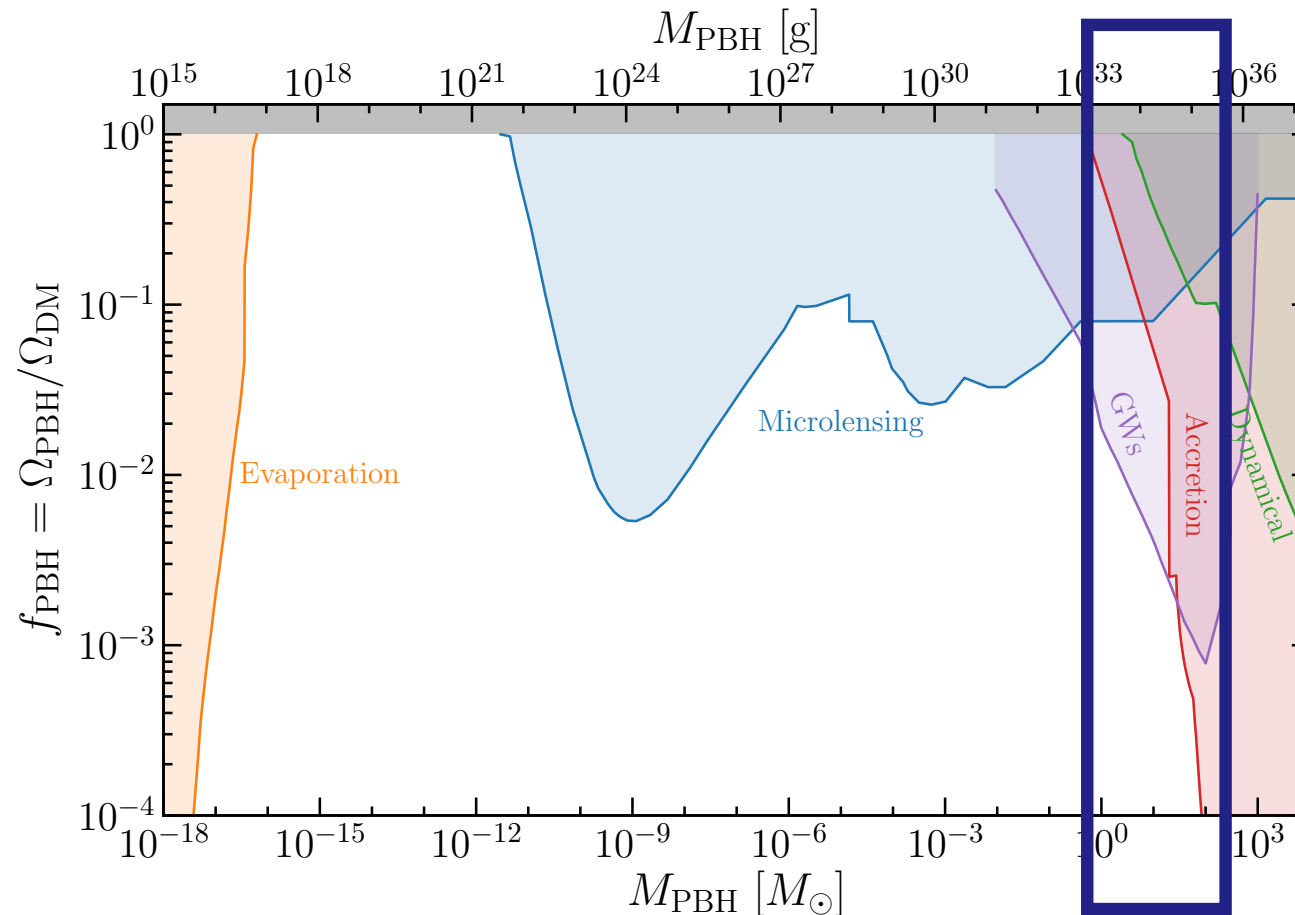


[github.com/
bradkav/
PBHbounds](https://github.com/bradkav/PBHbounds)

Primordial black holes (PBHs)

Zeldovich 67, Hawking MNRAS 152 (1971), Carr and Hawking MNRAS 168 (1974)

- ▶ BHs can form in early universe from *collapse of overdense regions* $\left(\frac{\delta\rho}{\rho} \gtrsim \frac{1}{3}\right)$
- ▶ Mass related to horizon size at time of formation



[github.com/
bradkav/
PBHbounds](https://github.com/bradkav/PBHbounds)

Outline

(1) Detecting black holes in the Milky Way

▶ *Gas accretion*

- Based on:
- ▶ F.Scarcella, D.Gaggero et al. 2012.10421 (MNRAS),
 - ▶ F.Scarcella, D.Gaggero, J. Garcia-Bellido PoS ICRC2021 (2021) 565

(2) Detecting black holes at cosmological distances

▶ *Gravitational waves*

- Based on:
- ▶ Master thesis, Cristina Fernandez
 - ▶ Master thesis, Tania Franco
 - ▶ F.Scarcella, D.Gaggero et al. 2205.02639



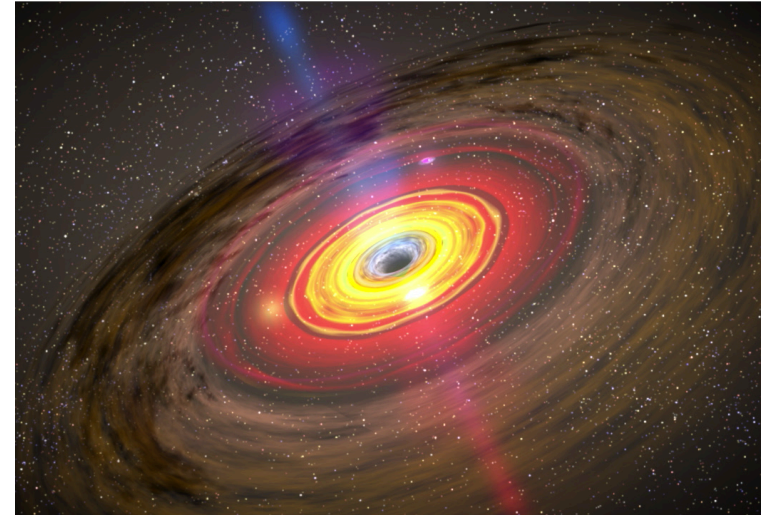
Gas accretion in the Milky Way

- Based on:
- 2012.10421 (MNRAS) with D. Gaggero, M. Ricotti, G. Bertone et al.
 - PoS ICRC2021 (2021) 565, 22xx.xxx with D. Gaggero, J. Garcia-Bellido

Detecting BHs in the Milky Way

BH expected to *accrete gas* from *dense clouds*

- ▶ Non-thermal radiation (X-ray/radio)
- ▶ Large population of PBHs + clouds at GC
- ▶ Channel used to set bound on PBH abundance [Manshanden et al. 1812.07967](#)
- ▶ No accreting isolated BH identified so far



Goals

- ➔ Robustness of bound wrt *astrophysical uncertainties*
- ➔ Dependence bound on the *PBH model (mass function)*

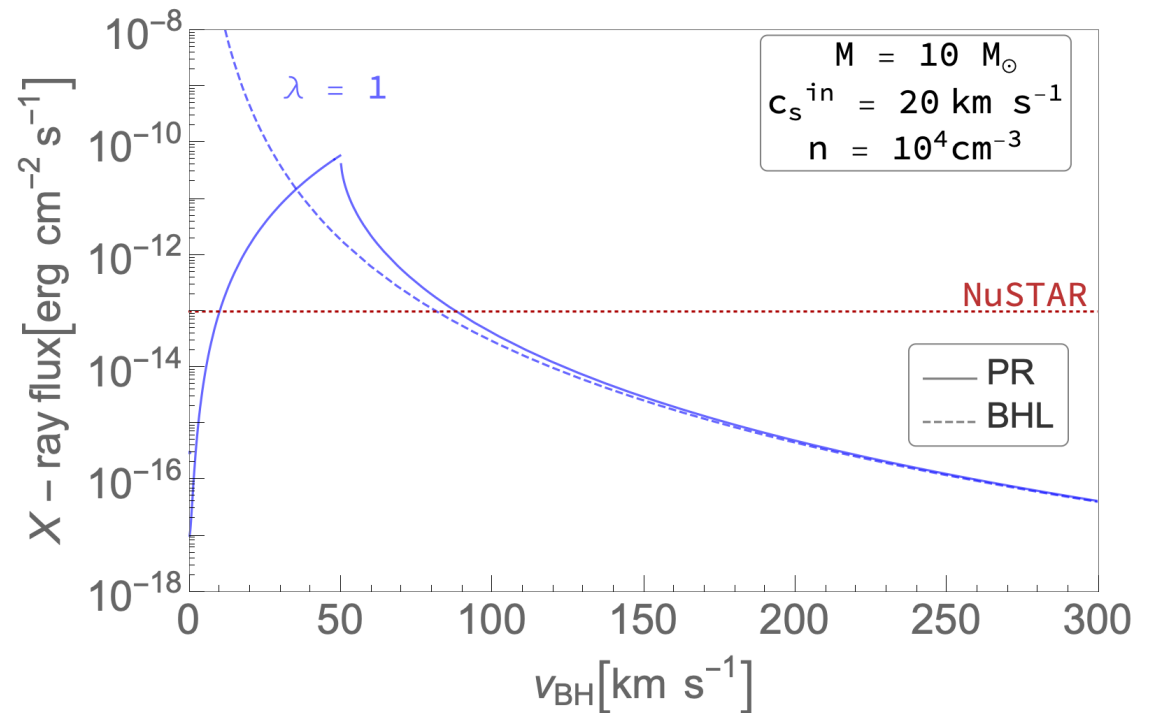
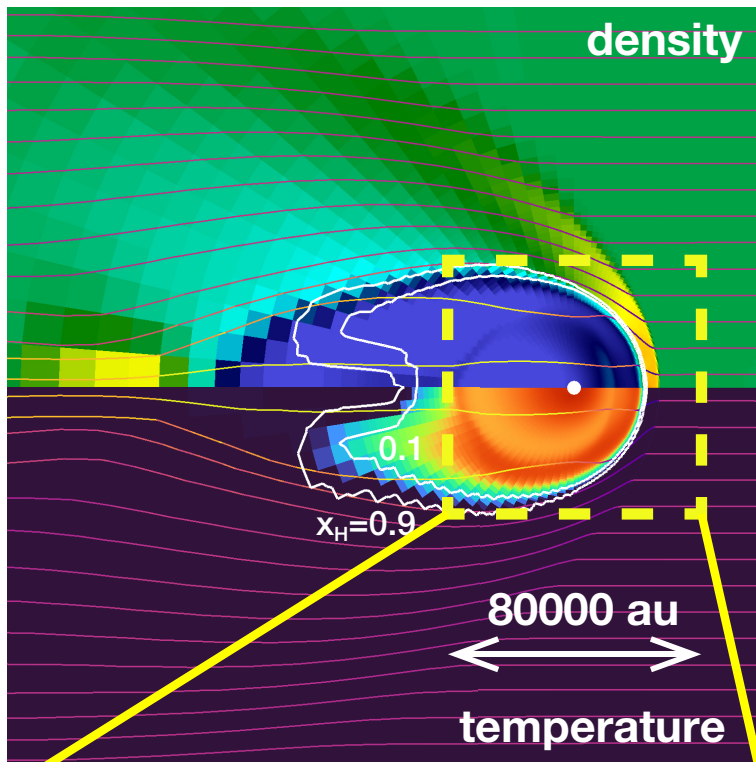
Modelling the accretion rate

$$L \propto \dot{M}^2$$

Park-Ricotti model, *radiative feedback*

211.0542, 2003.05625

$$\dot{M} = 4\pi \frac{(GM)^2 \rho_{\text{in}}}{(v_{\text{in}}^2 + c_{\text{s,in}}^2)^{3/2}}$$



- ➔ Main variables $M, v_{\text{rel}}, n, c_{\text{s}}^{\text{in}}$
- ➔ Threshold effect

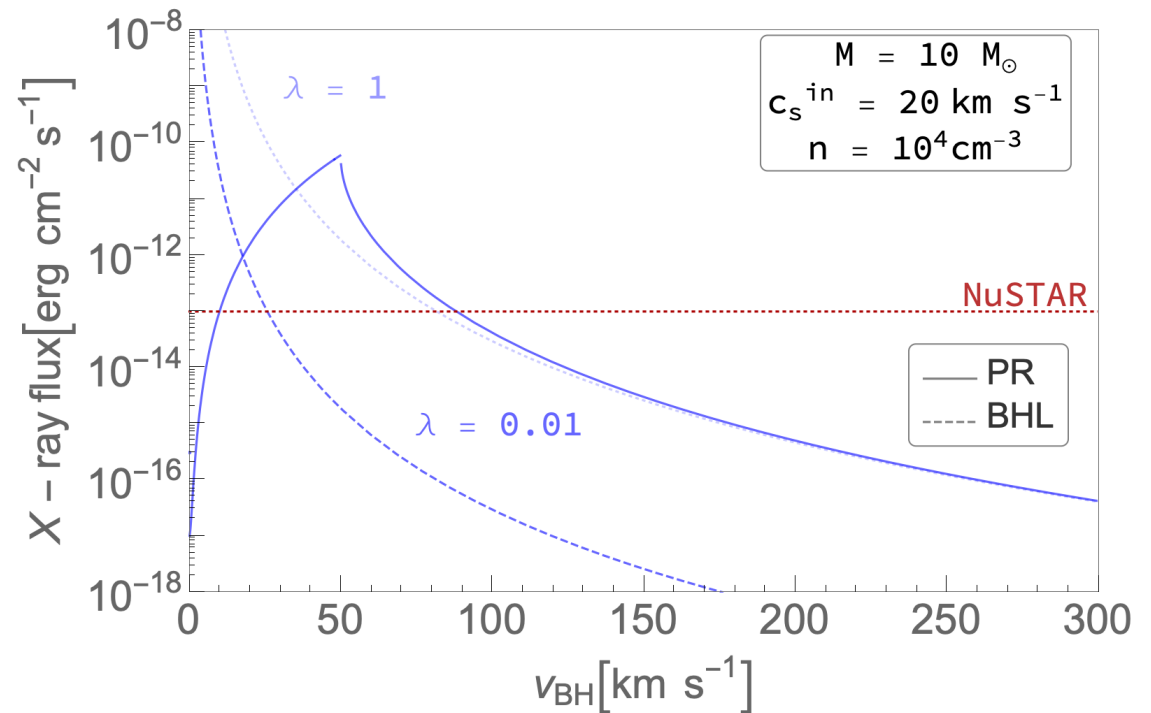
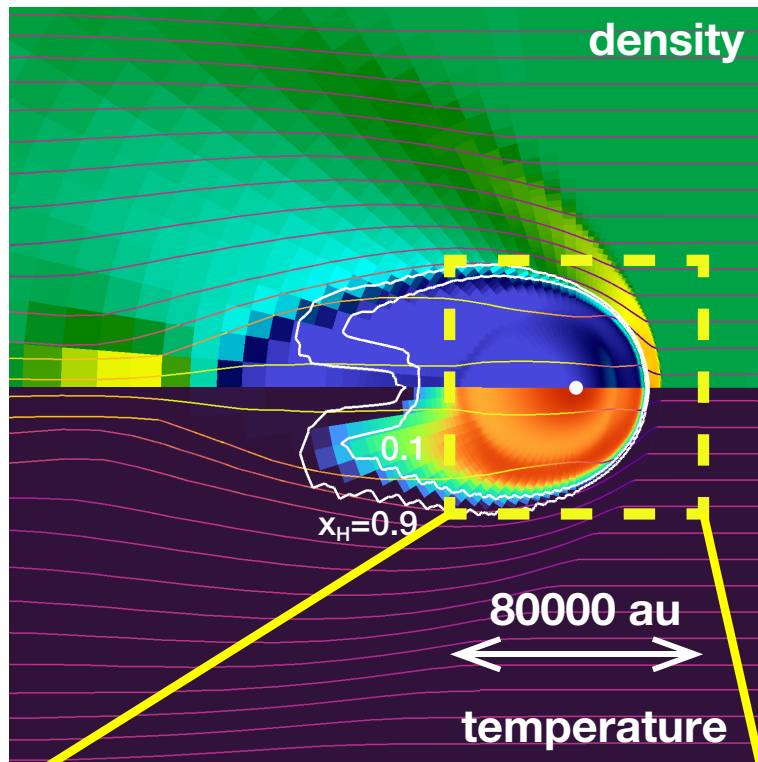
Modelling the accretion rate

$$L \propto \dot{M}^2$$

Park-Ricotti model, *radiative feedback*

211.0542, 2003.05625

$$\dot{M} = 4\pi \frac{(GM)^2 \rho_{\text{in}}}{(v_{\text{in}}^2 + c_{\text{s,in}}^2)^{3/2}}$$



- ➔ Main variables $M, v_{\text{rel}}, n, c_{\text{s}}^{\text{in}}$
- ➔ Threshold effect

Expected number of visible sources

Semi-analytic study of impact of uncertainties on the predictions

$$N^{\text{sources}}(\phi^*) = N^{\text{tot}} \int_{\phi(v_{\text{BH}}, M, d, \{p_i\}) > \phi^*} P(v_{\text{BH}}) P(M) P(r) P(n) dv_{\text{BH}} dM dr dn$$

Modelling the *BH population*:

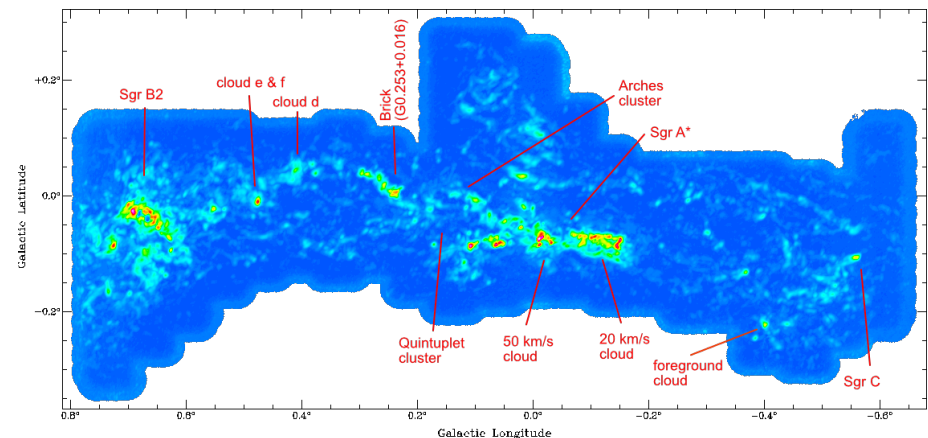
- ▶ Spatial distribution (NFW)
- ▶ Velocity distribution (MB)
- ▶ Mass distribution

- ▶ Temperature of ionised 'bubble' (c_s^{in})

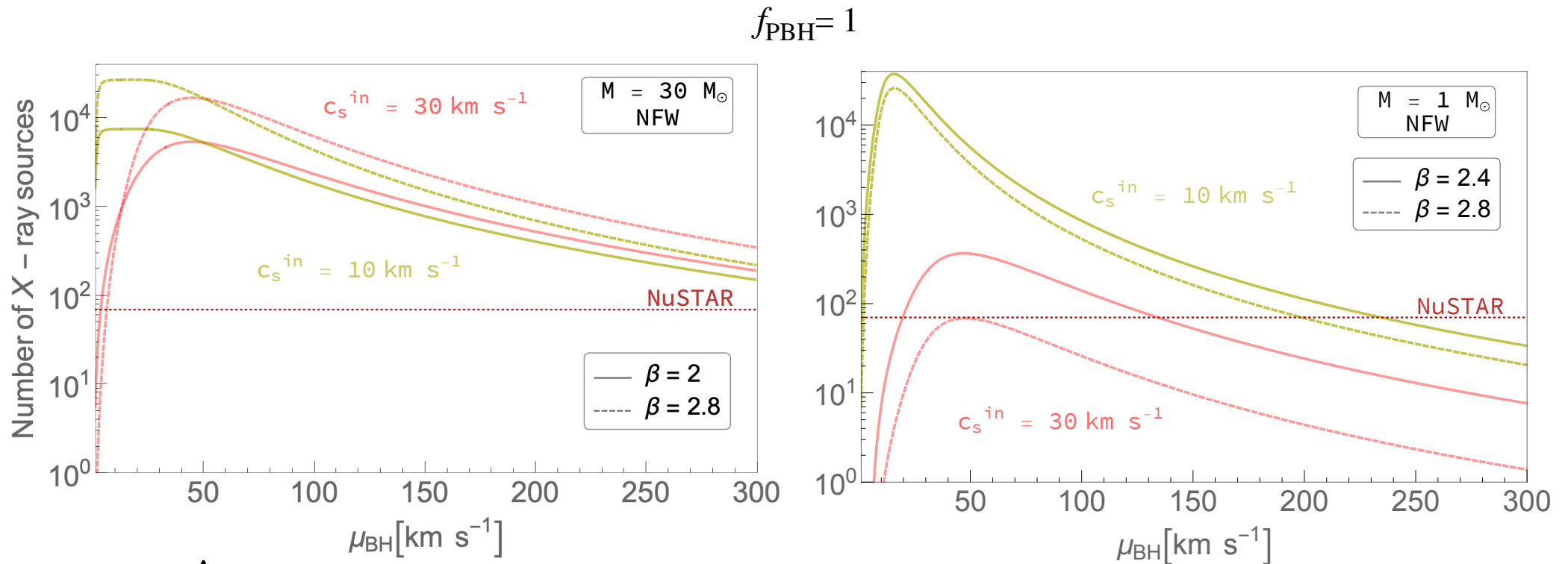
Modelling the *molecular clouds*:

- ▶ Density distribution $P(n) \propto n^{-\beta}$

Ferrière et al: 0702532

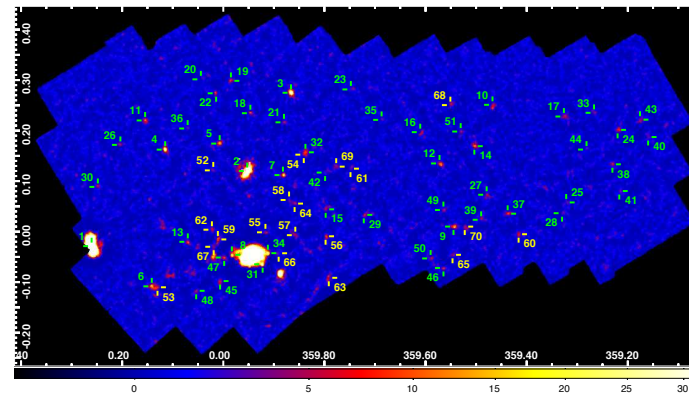


Results: number of X-ray sources (monochromatic)



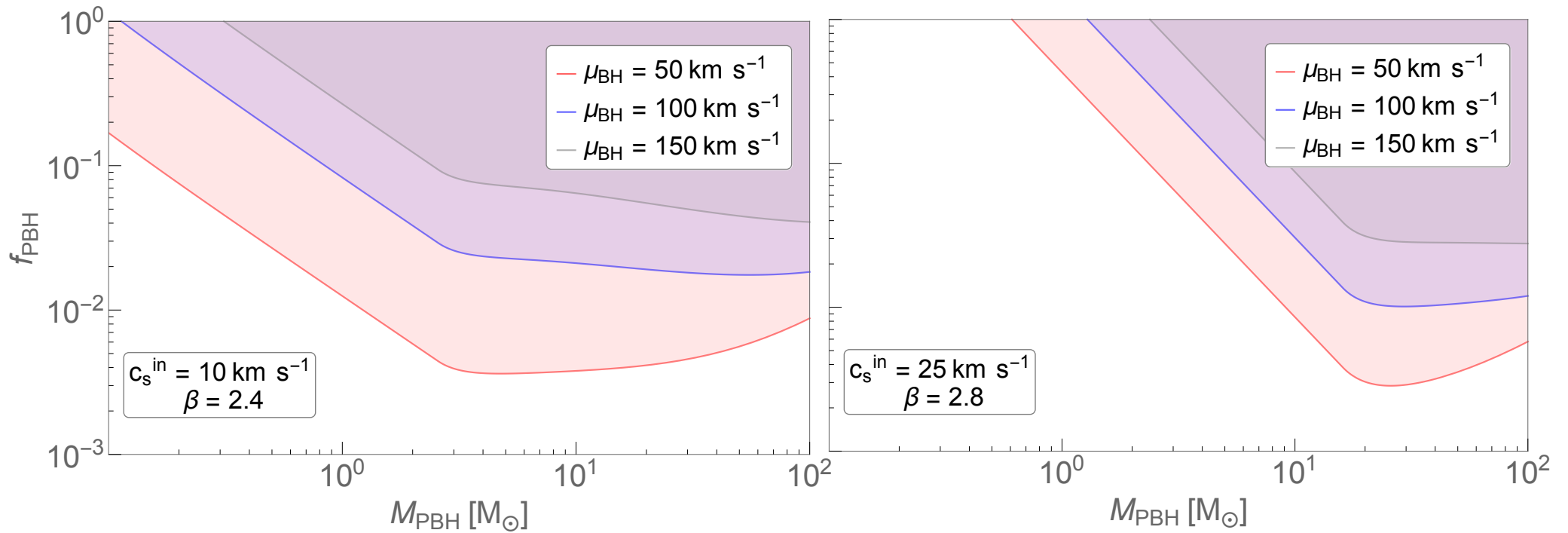
↑
Eddington
inversion
formula

NuSTAR survey: 3-40 keV, ~70 sources GC, 1605.03882



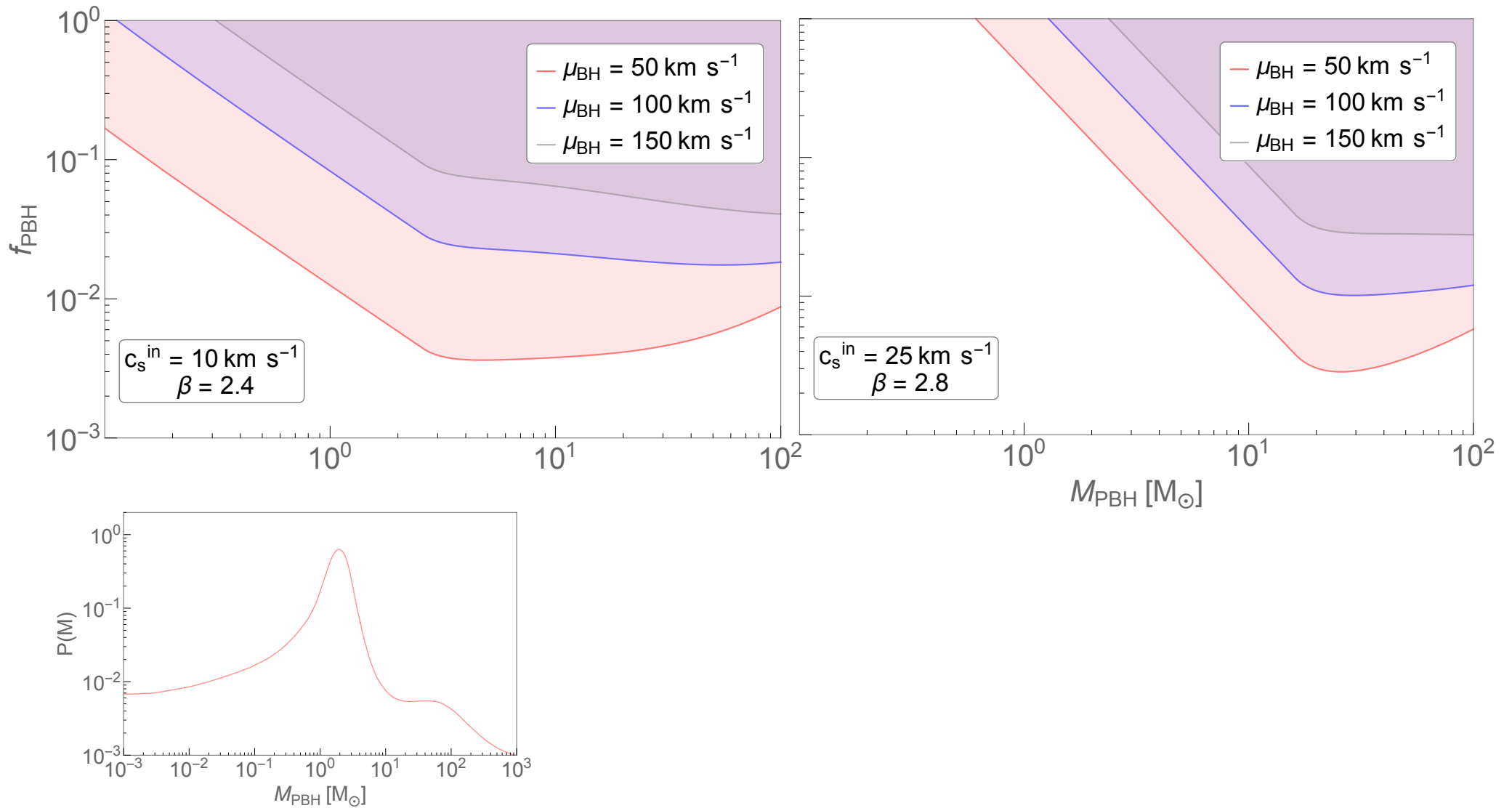
Results: abundance constraints

NuSTAR survey: 3-40 keV, ~70 sources GC, 1605.03882



Results: abundance constraints

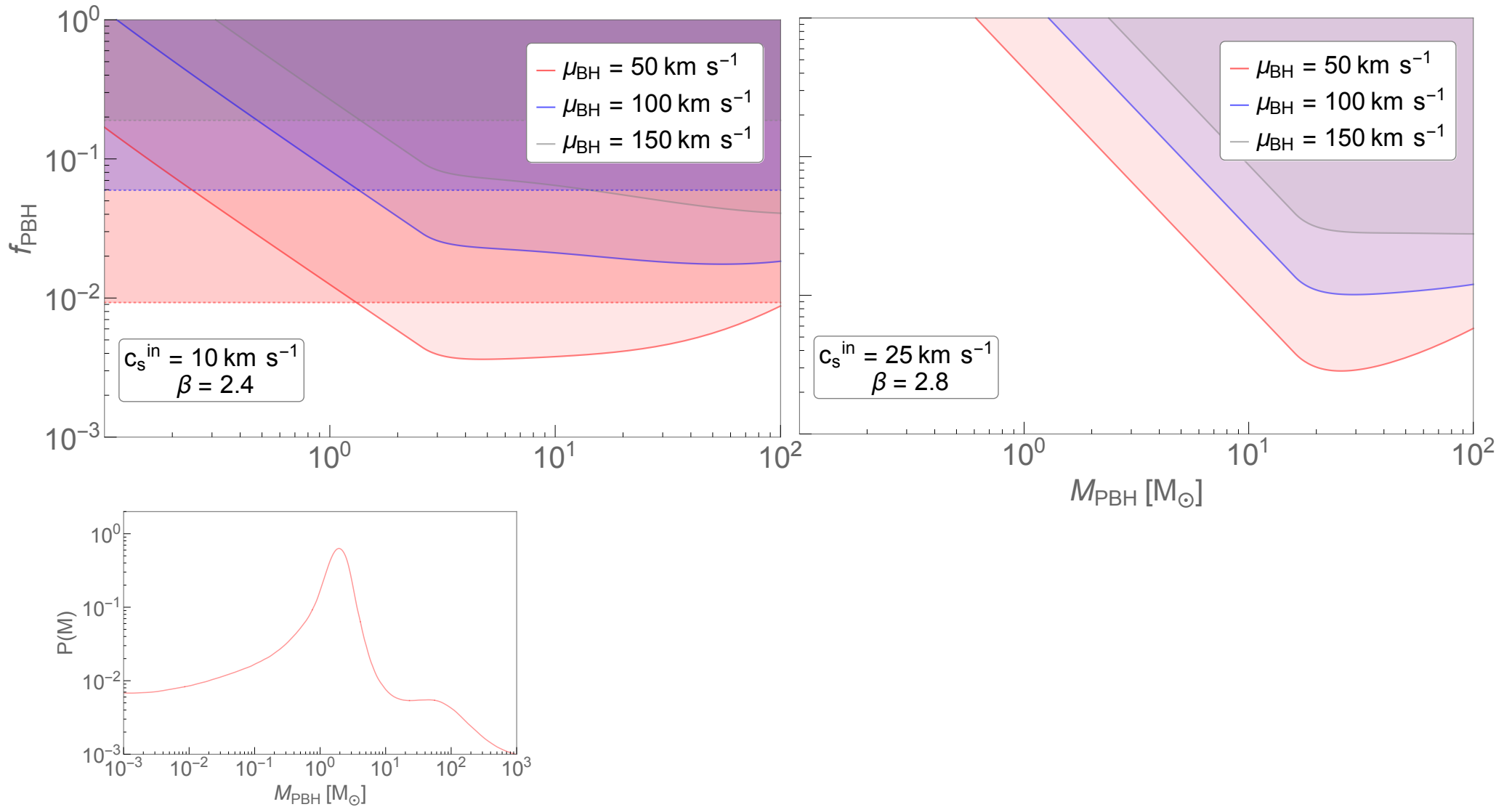
NuSTAR survey: 3-40 keV, ~70 sources GC, 1605.03882



Jedamzik 1996, Byrnes et al. 2018, Carr et al. 2019

Results: abundance constraints

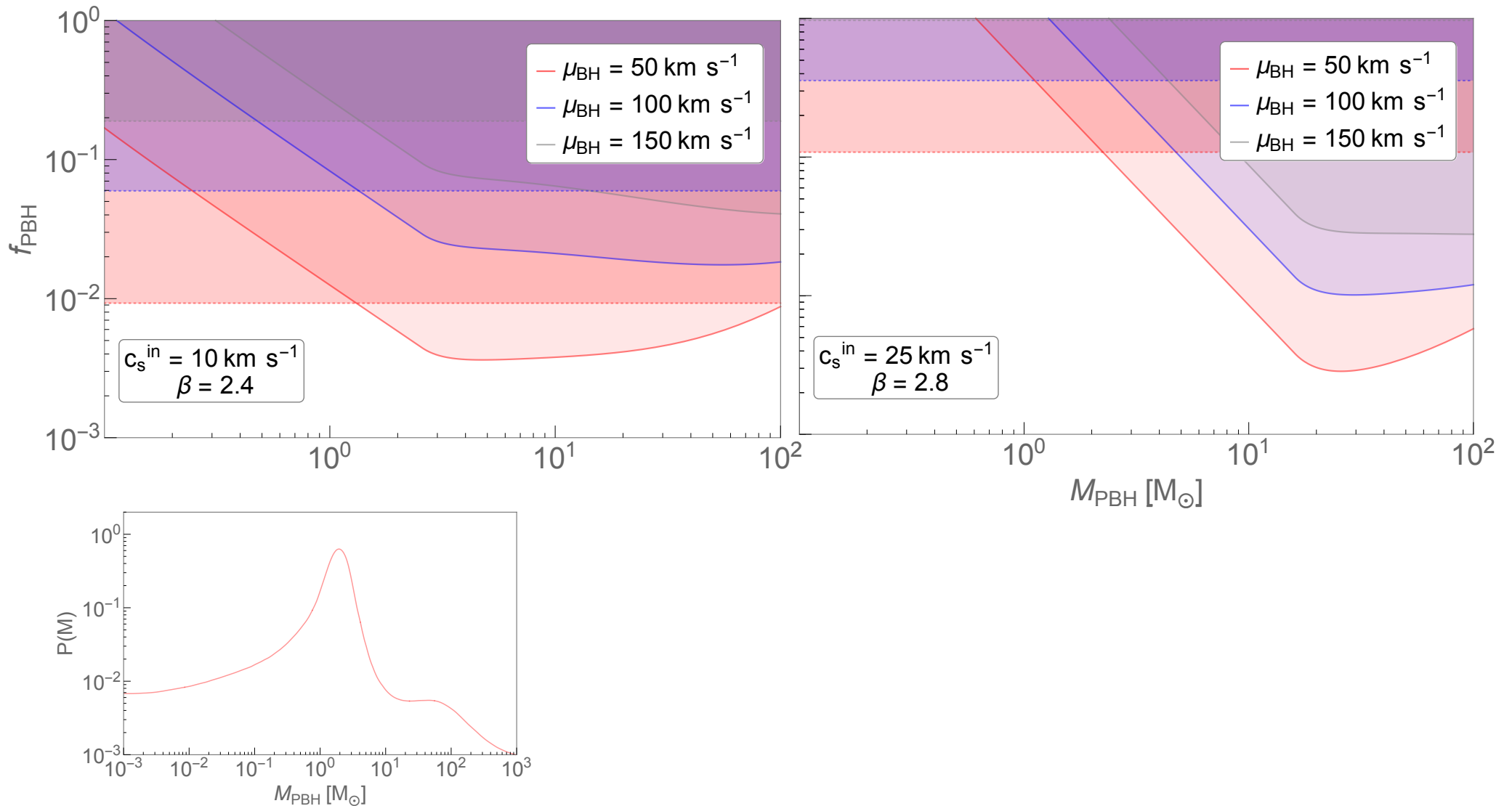
NuSTAR survey: 3-40 keV, ~70 sources GC, 1605.03882



Jedamzik 1996, Byrnes et al. 2018, Carr et al. 2019

Results: abundance constraints

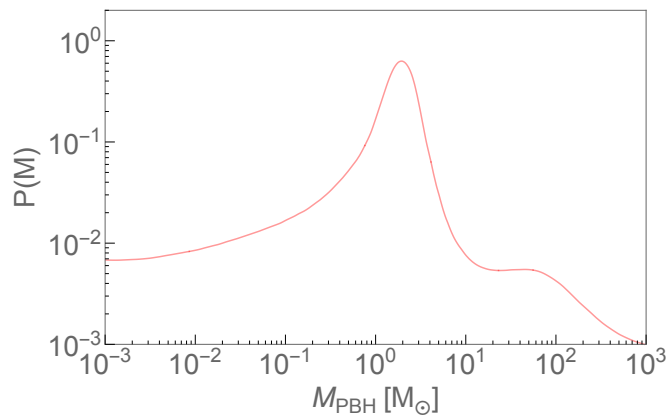
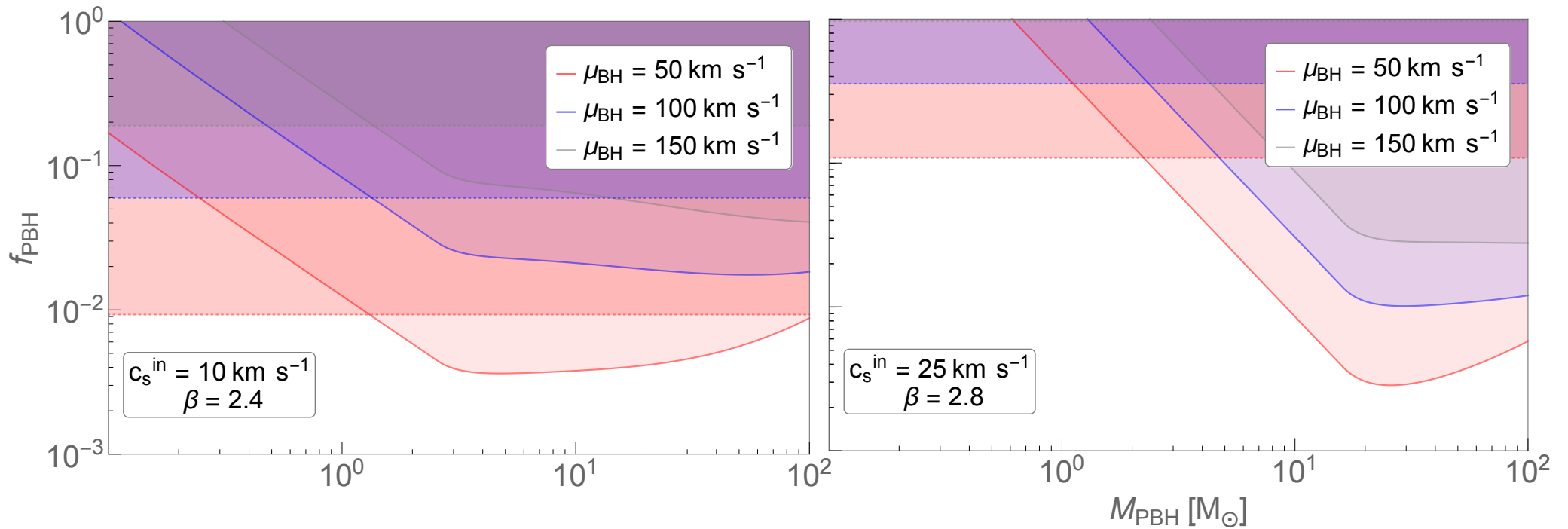
NuSTAR survey: 3-40 keV, ~70 sources GC, 1605.03882



Jedamzik 1996, Byrnes et al. 2018, Carr et al. 2019

Results: abundance constraints

NuSTAR survey: 3-40 keV, ~70 sources GC, 1605.03882



➔ Dominated by $1 M_{\odot}$ peak: large uncertainties

➔ In progress: Bayesian posterior on f_{PBH}

PoS ICRC2021 (2021) 565

Jedamzik 1996, Byrnes et al. 2018, Carr et al. 2019

(1) Summary

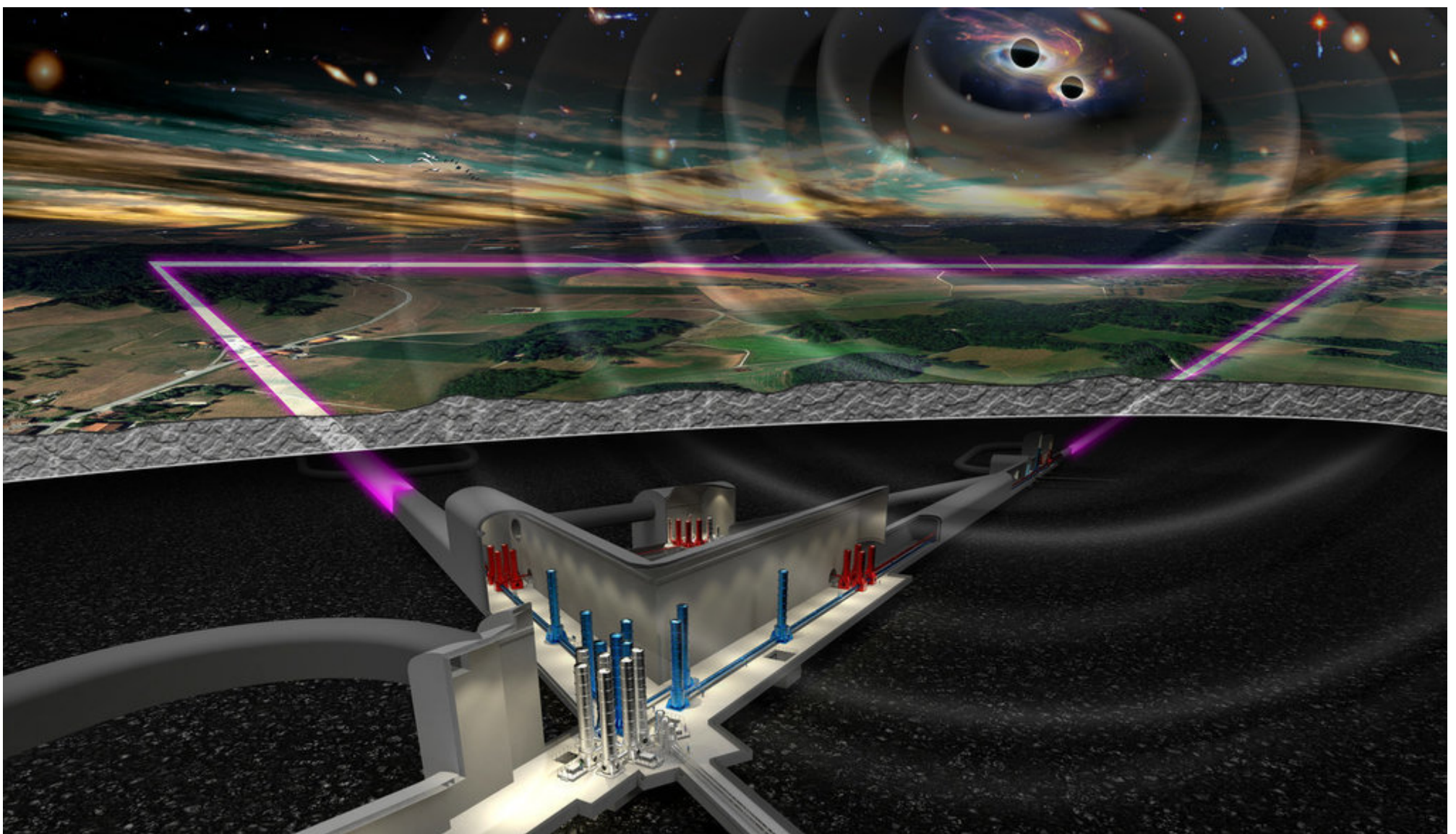
- ➔ Useful channel to *constrain PBH abundance above* $\sim 10 M_{\odot}$
- ➔ *Large uncertainties* for PBH mass $\sim 1 M_{\odot}$
- ➔ Uncertainties *reflected on multimodal mass function*

Caveats

- ➔ PBH *spatial distribution* (correlation with cloud positions, *clusters*)
- ➔ PBH velocity dispersion

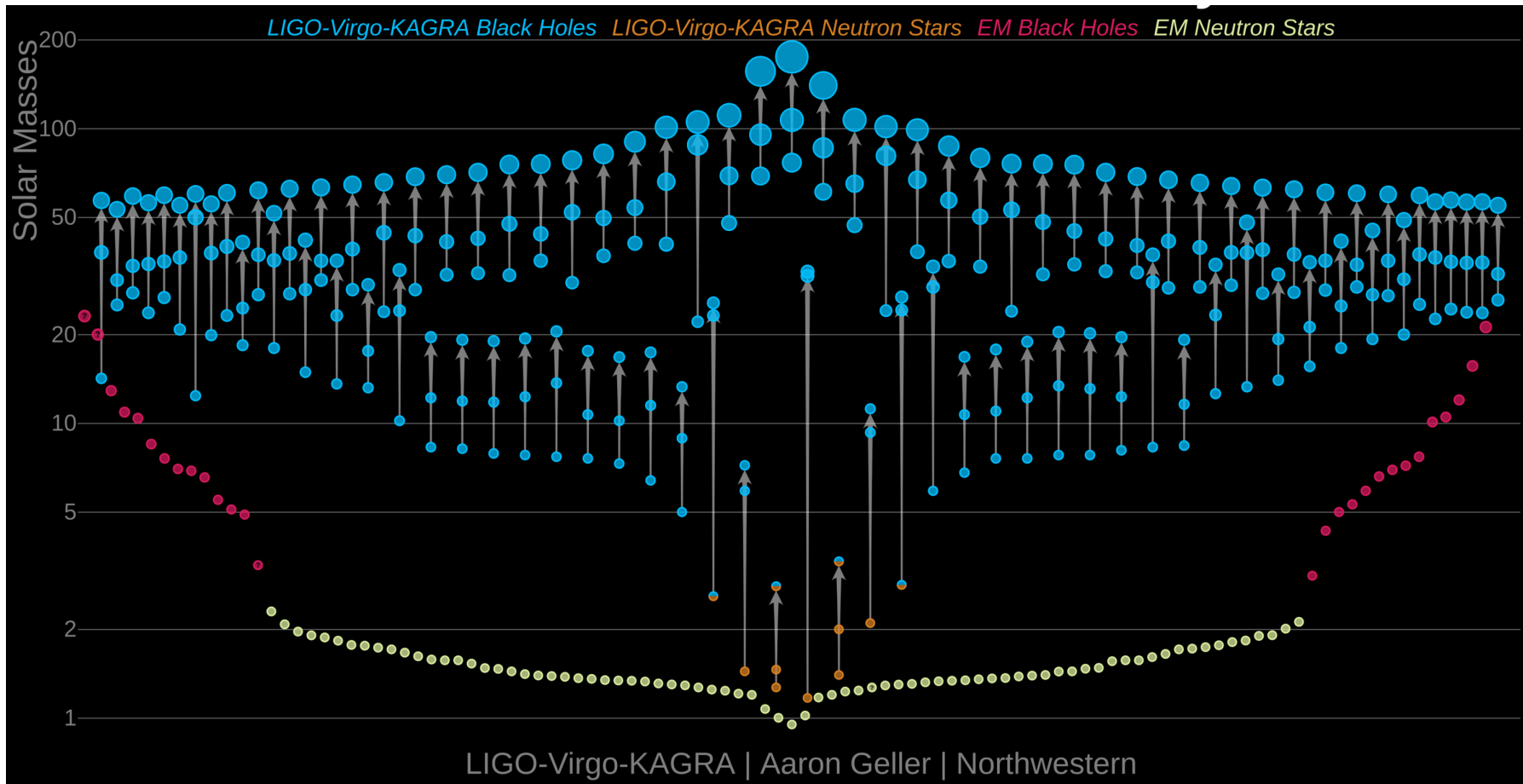
Prospects

- ➔ Identification of sources as BH - *dedicated multi-wavelength analysis*
- ➔ Identification of sources as *PBH or ABH - challenging*

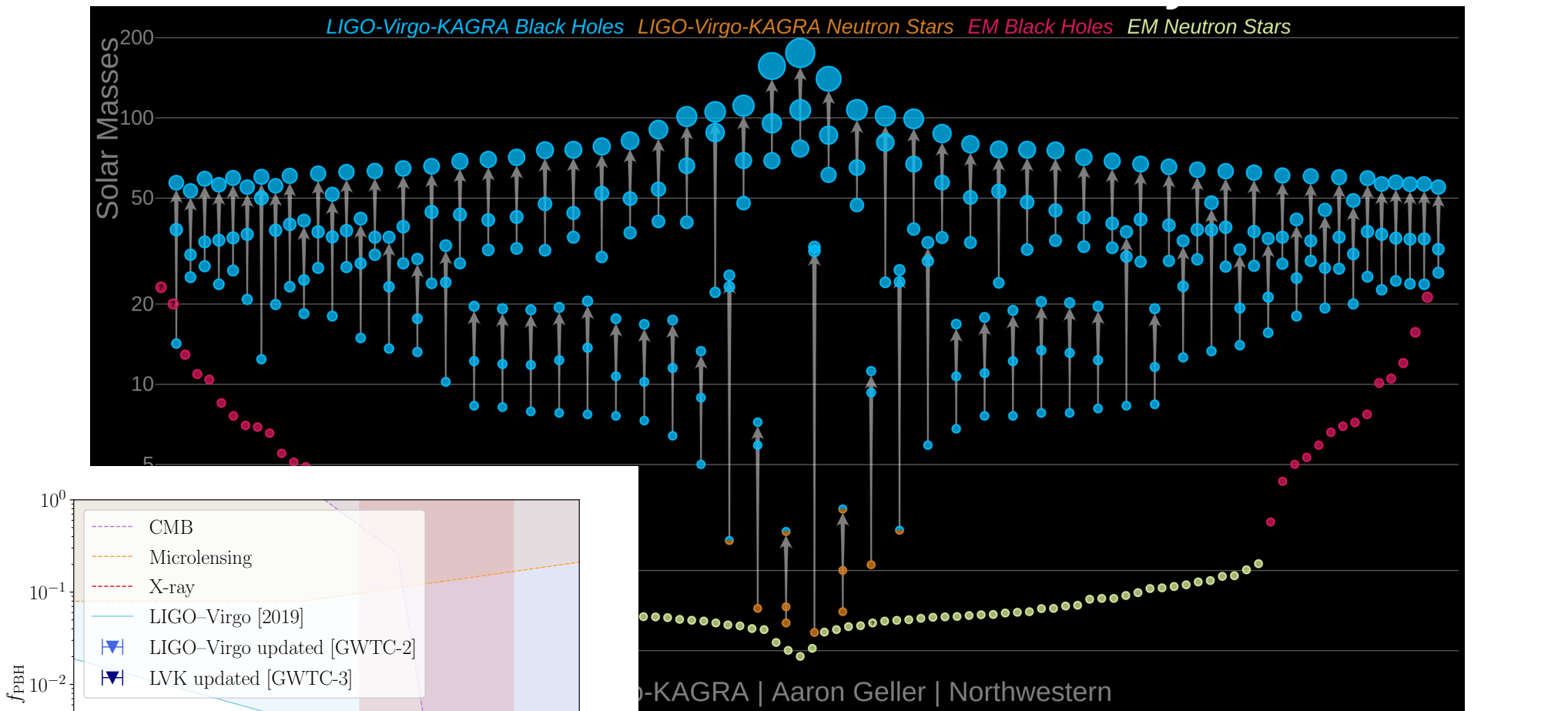


Black hole mergers at high redshift

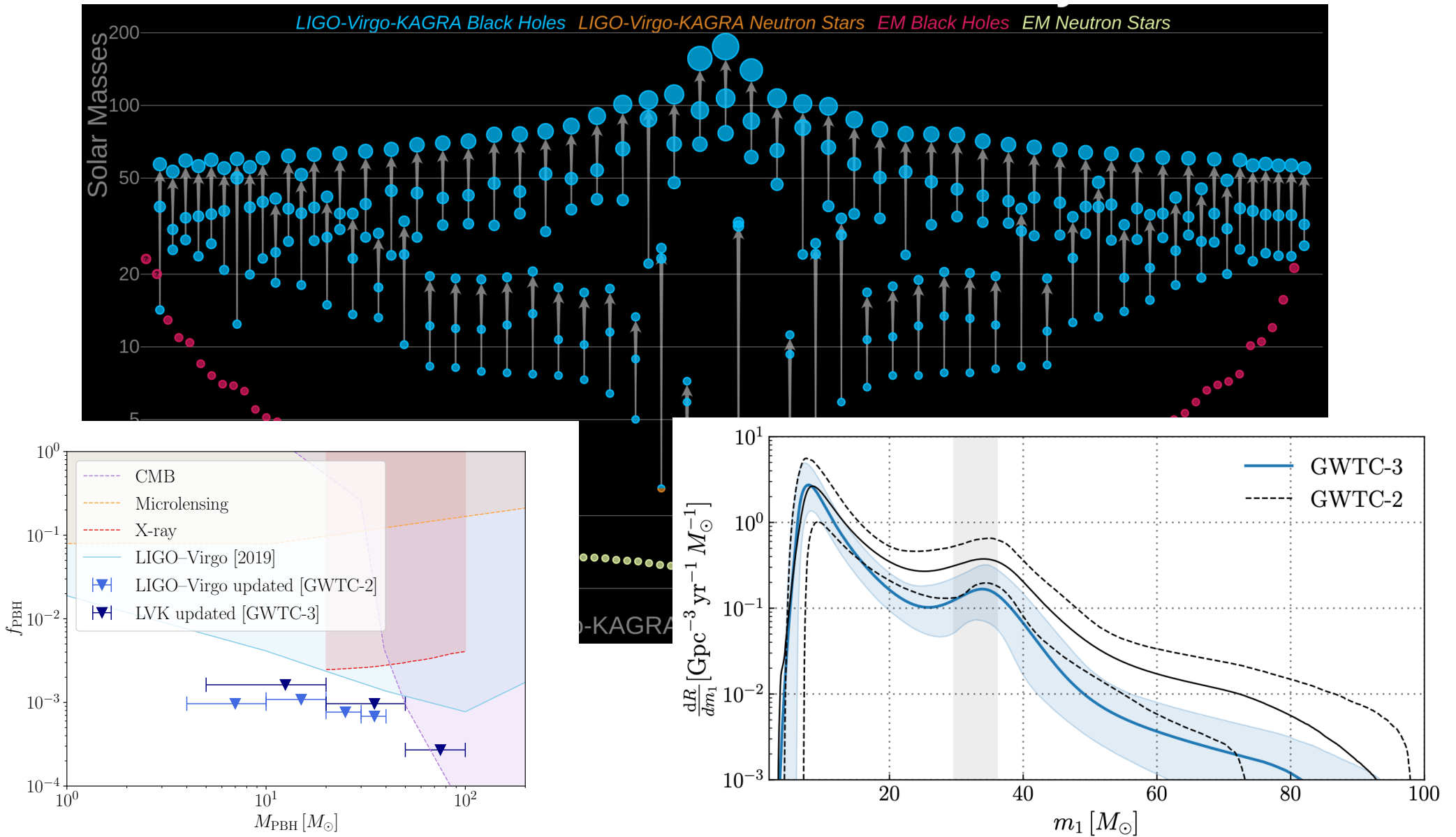
Based on 2205.02639 with M. Martinelli, N. Hogg, P. Fleury, D. Gaggero, B. Kavanagh

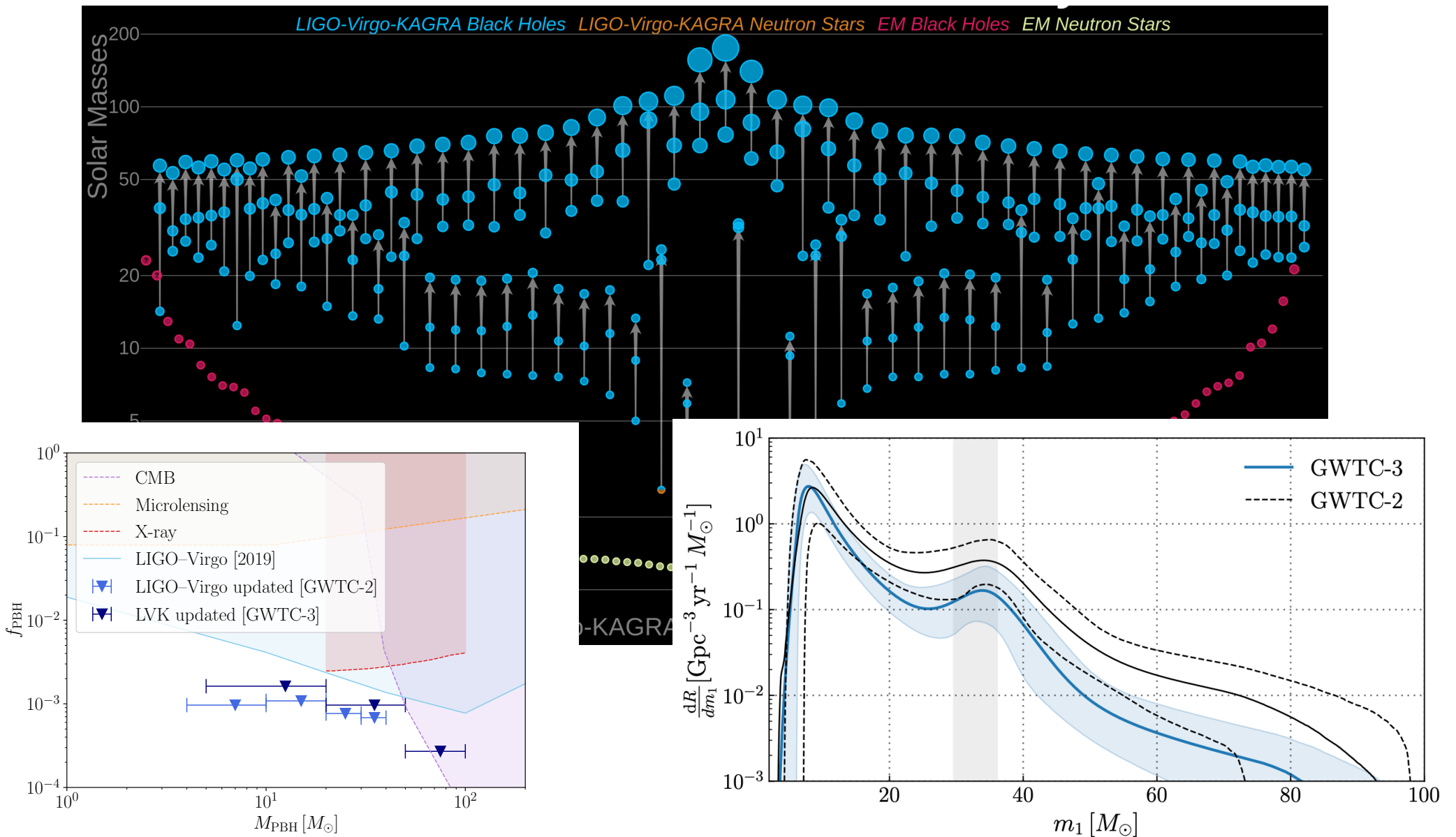


GWTC-3 Catalog



GWTC-3 Catalog

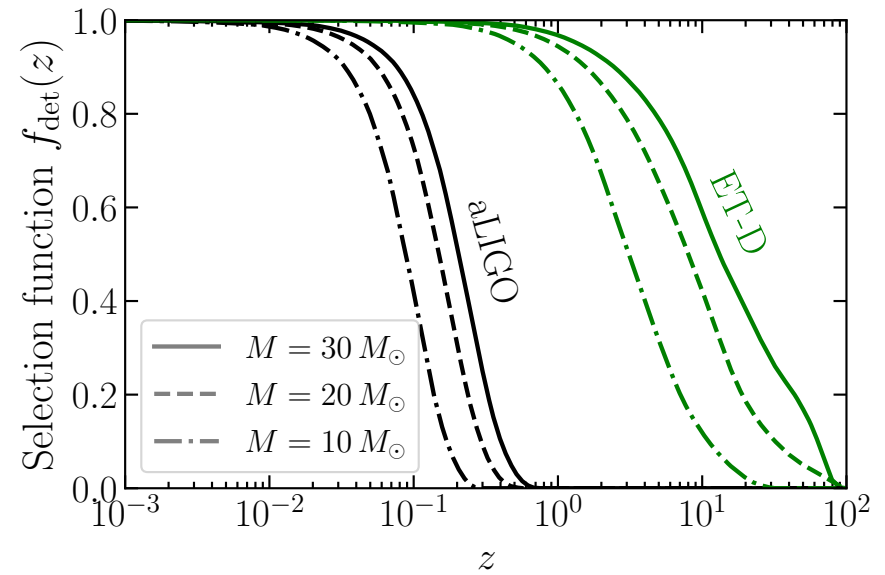
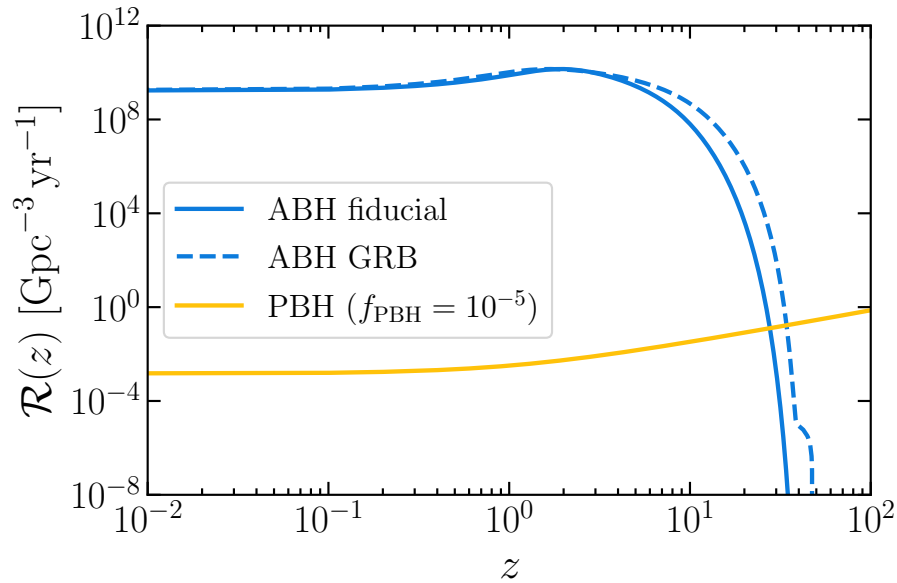




Clean channel: high redshift (*no astrophysical* background)

Prospects for the Einstein Telescope

Can we identify primordial black holes with future gravitational wave observatories?

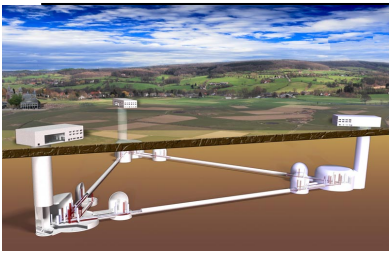


Forecast for Einstein telescope to asses:

- ▶ ability to *detect* PBH
- ▶ ability to *measure* PBH abundance

$$R_{\text{det}}(z) = f_{\text{det}}(z) \frac{\mathcal{R}(z)}{1+z} \frac{dV_c}{dz}$$

➔ **Simulate** PBH merger events and *detector's response*



ET mock data generation

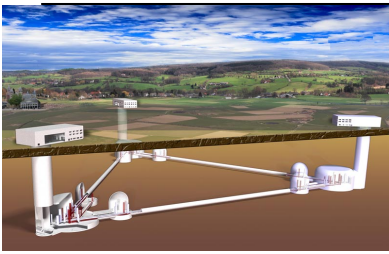
gitlab.com/matmartinelli/darksirens

Sketch of the *mock data generation* algorithm:

- ▶ Compute expected number of events (T_{obs})
- ▶ Each event (redshift, position, inclination) \rightarrow *waveform* (PyCBC)
- ▶ ET antenna patterns \rightarrow strain $h(f)$
- ▶ Compute *signal-to-noise ratio* ρ_i
- ▶ Discard faint events ($\rho_i < 8$)
- ▶ Draw ΔD_i *instrumental error on distance* from Gaussian
- ▶ Extract observed value of D_L
- ▶ Obtain error on D_L including lensing effects

$$\rho_i = \left[4 \int_{f_{\text{lower}}}^{f_{\text{upper}}} df \frac{h_i(f)h_i^*(f)}{S_n(f)} \right]^{\frac{1}{2}}$$

$$\sigma_i^{\text{inst}} = 2\tilde{D}_i/\rho_i$$



ET mock data generation

gitlab.com/matmartinelli/darksirens

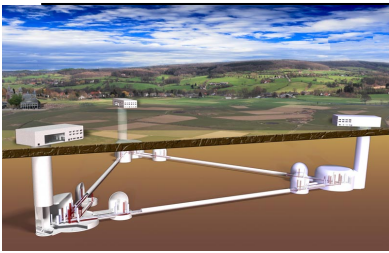
Sketch of the *mock data generation* algorithm:

- ▶ Compute expected number of events (T_{obs})
- ▶ Each event (redshift, position, inclination) \rightarrow *waveform* (PyCBC)
- ▶ ET antenna patterns \rightarrow strain $h(f)$
- ▶ Compute *signal-to-noise ratio* ρ_i
- ▶ Discard faint events ($\rho_i < 8$)
- ▶ Draw ΔD_i *instrumental error on distance* from Gaussian
- ▶ Extract observed value of D_L
- ▶ Obtain error on D_L including lensing effects

$$\rho_i = \left[4 \int_{f_{\text{lower}}}^{f_{\text{upper}}} df \frac{h_i(f)h_i^*(f)}{S_n(f)} \right]^{\frac{1}{2}}$$

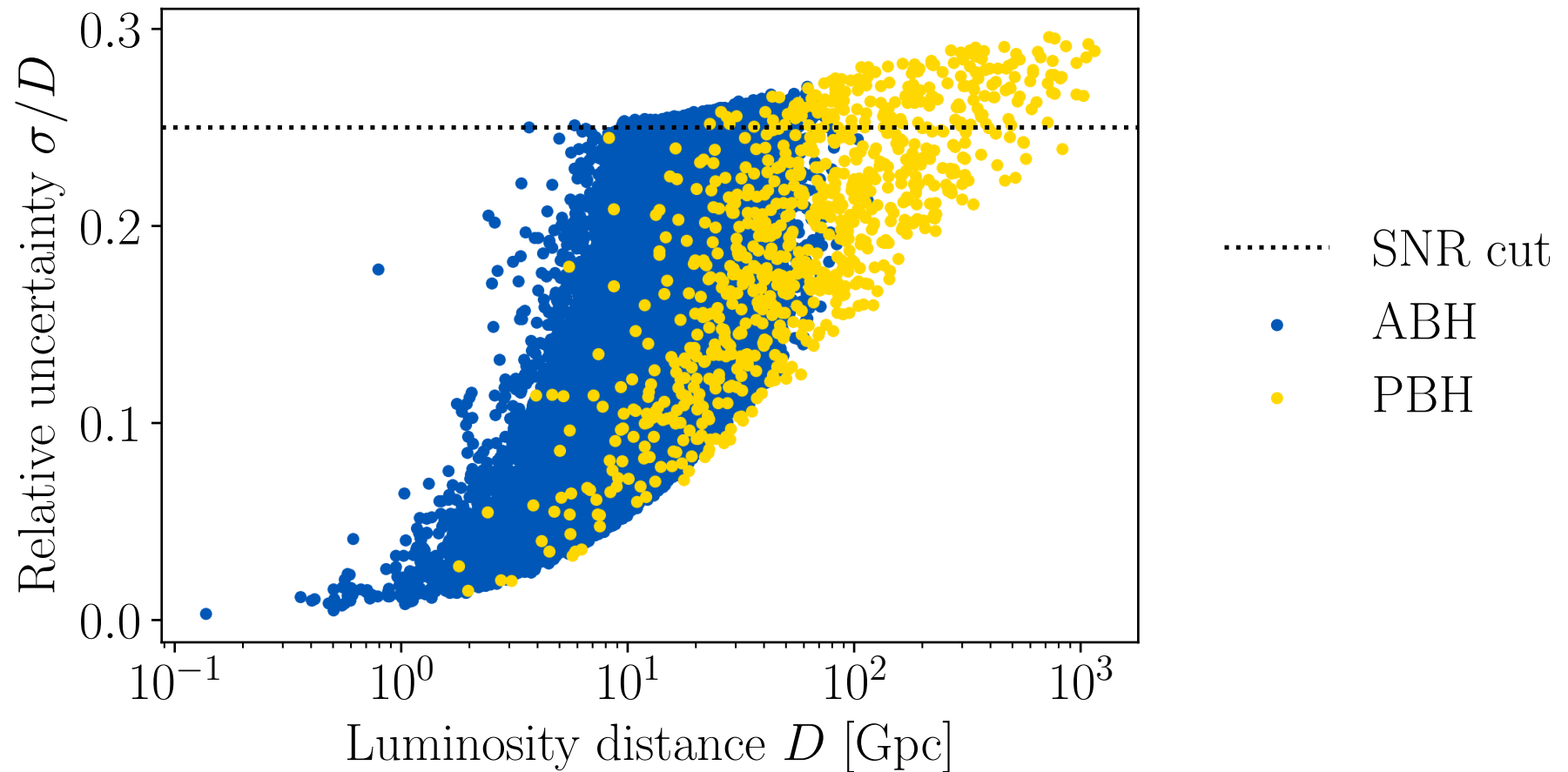
$$\sigma_i^{\text{inst}} = 2\tilde{D}_i/\rho_i$$

Final result: *mock catalog* (D_i, σ_i)



ET mock data generation

gitlab.com/matmartinelli/darksirens

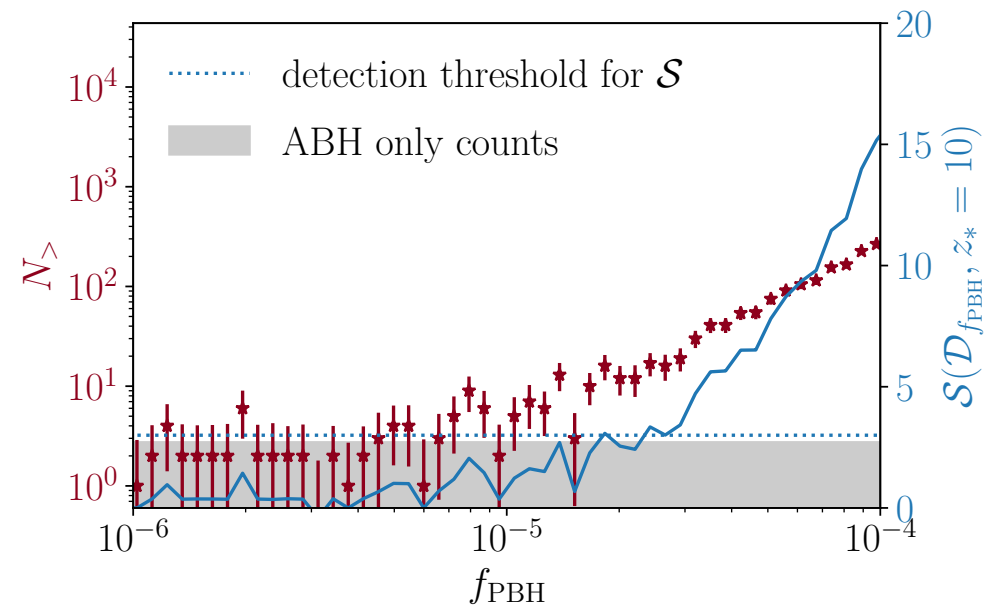
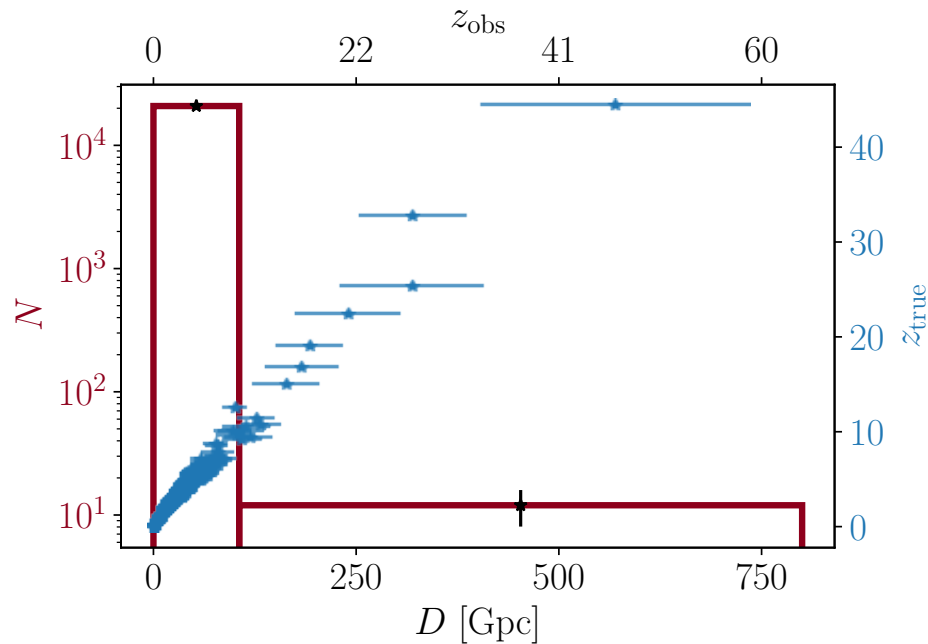


Final result: *mock catalog* (D_i, σ_i)

Data analysis - 1

“Cut-and-count”

- ▶ *Divide data in two bins*, evaluate $N_{>}$: # events with $z > z^*$
- ▶ Generate catalogs for *different values of* f_{PBH} , evaluate $N_{>}$
- ▶ Compare with null hypothesis: *ABH only* data set $\rightarrow N_{>} = 1 \pm 1.7$



➔ *Smallest detectable fraction* (3σ): $f_{\text{PBH}} \approx 10^{-5} \rightarrow N_{>} = 16 \pm 5$

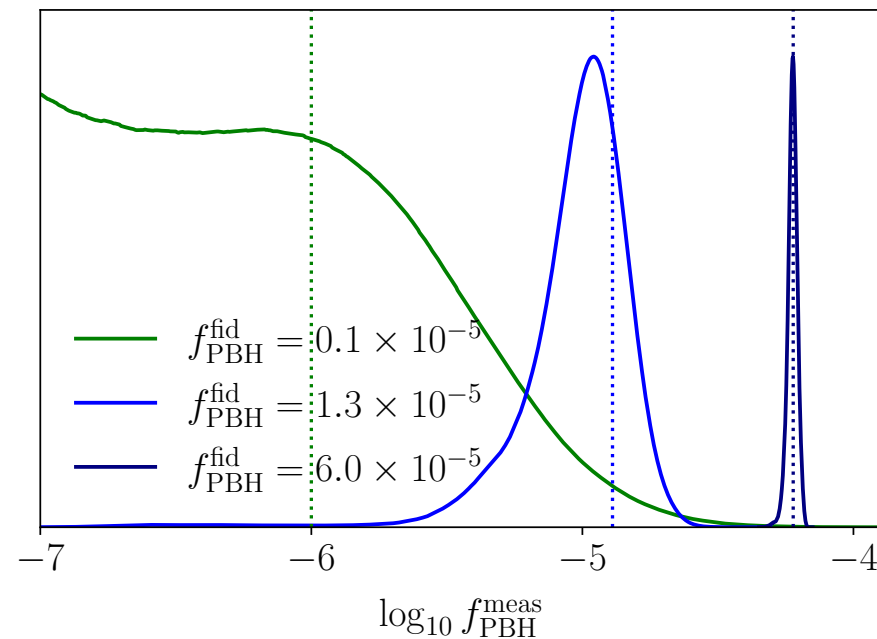
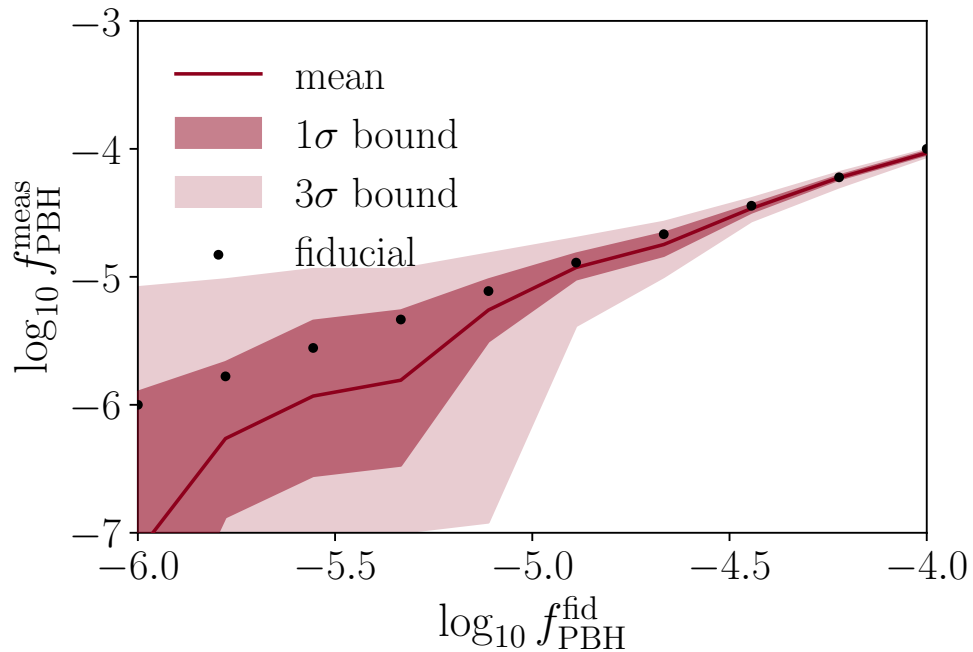
Data analysis - 2

- ▶ **Unbinned likelihood** - probability of a set of observed events

$$\mathcal{L}(f_{\text{PBH}} | \mathcal{D}) = \frac{\bar{N}_{\text{obs}}(f_{\text{PBH}})^{N_{\text{obs}}} e^{-\bar{N}_{\text{obs}}(f_{\text{PBH}})}}{N_{\text{obs}}!} \times \prod_{i=1, N_{\text{obs}}} p(D_i | f_{\text{PBH}})$$

- ▶ **Posterior distribution for f_{PBH}**

$$p(f_{\text{PBH}} | \mathcal{D}) \propto \mathcal{L}(f_{\text{PBH}} | \mathcal{D}) \text{Pr}(f_{\text{PBH}})$$



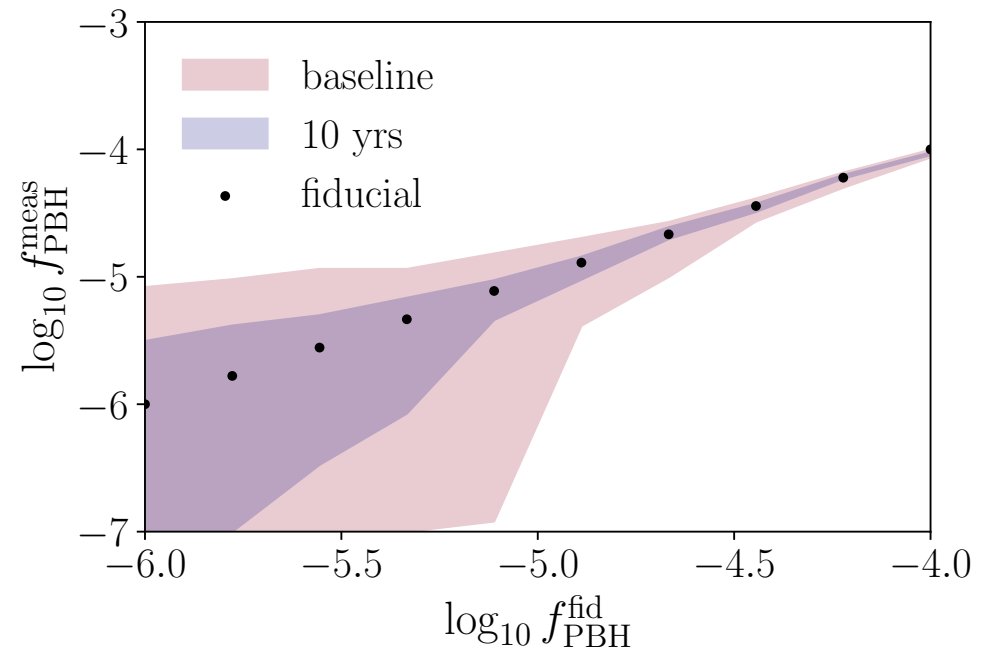
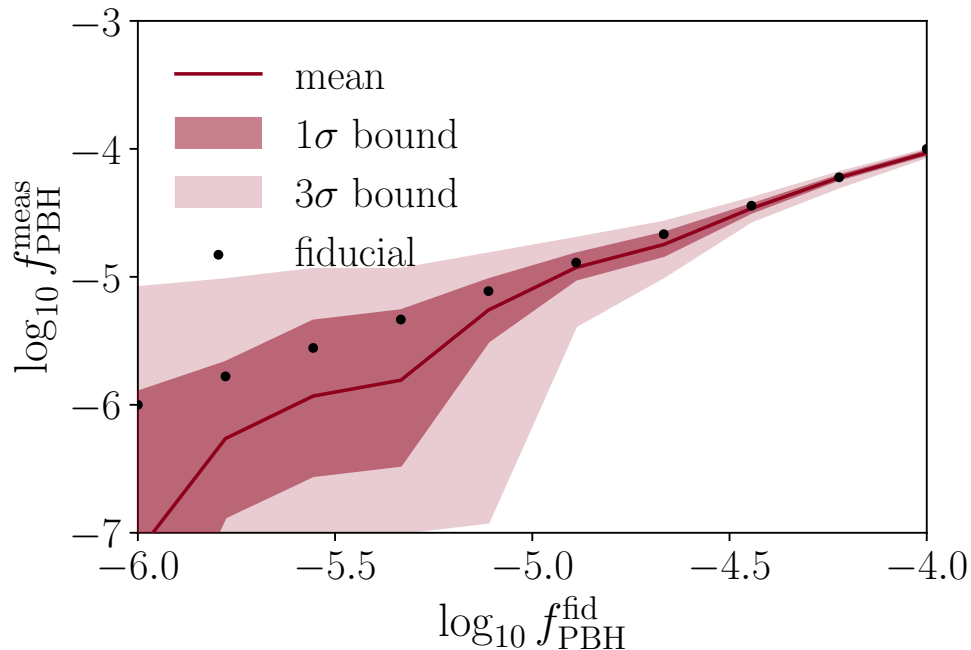
Data analysis - 2

- ▶ *Unbinned likelihood* - probability of a set of observed events

$$\mathcal{L}(f_{\text{PBH}} | \mathcal{D}) = \frac{\bar{N}_{\text{obs}}(f_{\text{PBH}})^{N_{\text{obs}}} e^{-\bar{N}_{\text{obs}}(f_{\text{PBH}})}}{N_{\text{obs}}!} \times \prod_{i=1, N_{\text{obs}}} p(D_i | f_{\text{PBH}})$$

- ▶ *Posterior distribution for f_{PBH}*

$$p(f_{\text{PBH}} | \mathcal{D}) \propto \mathcal{L}(f_{\text{PBH}} | \mathcal{D}) \text{Pr}(f_{\text{PBH}})$$



(2) Summary

- ➔ *Future GW observatories: powerful tool* to identify PBH signal
- ➔ Measure PBHs in *abundance as small as* $f_{\text{PBH}} \approx 10^{-5}$

Caveats

- ➔ *Signal* modelling (mass function, initial clustering, ...)
- ➔ Astrophysical *background* (pop III stars)

- ➔ Provided a *public tool* for generation of GW mock data

gitlab.com/matmartinelli/darksirens



Thank you