

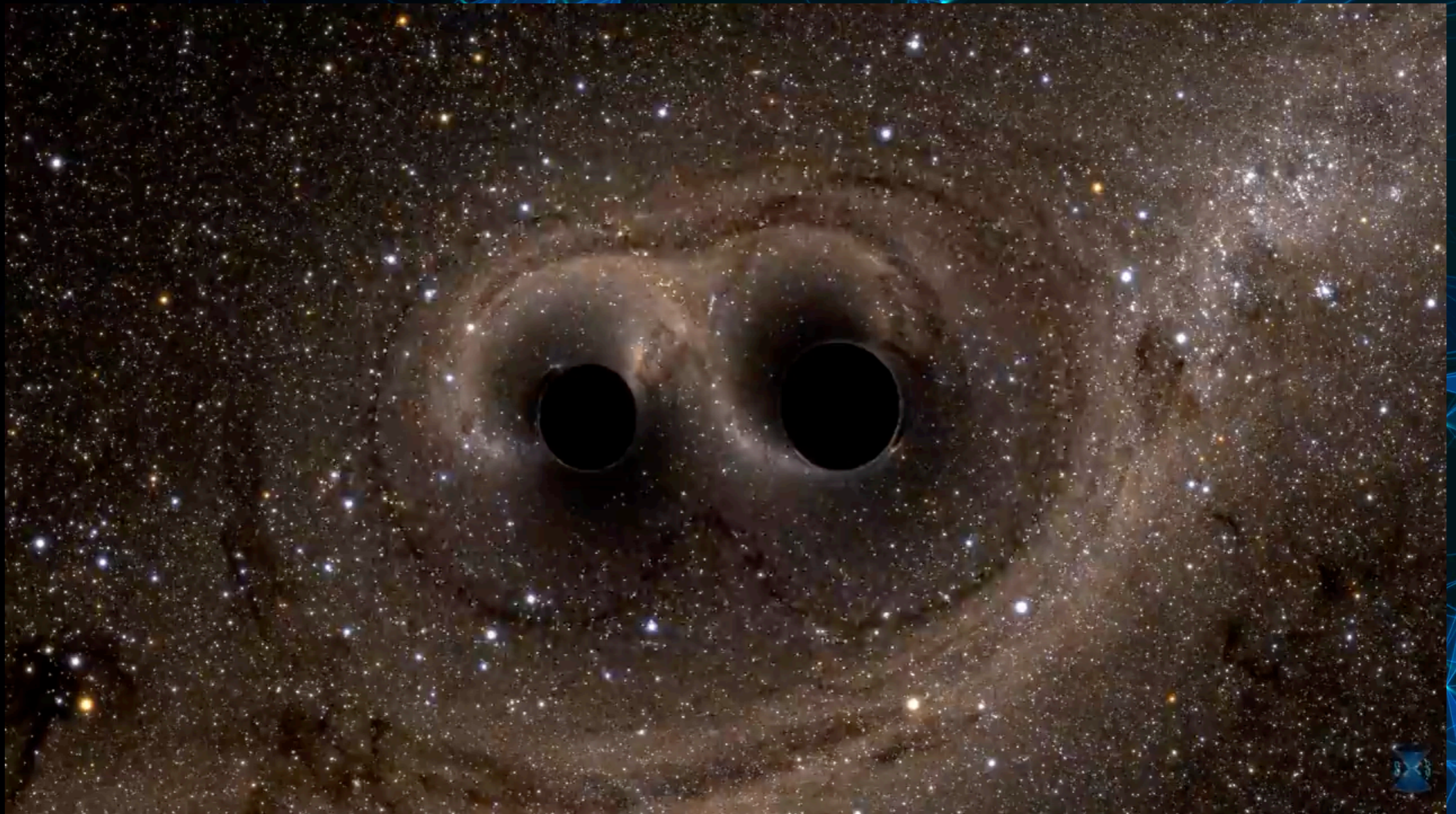
# ACTIVE LEARNING FOR GW MODELLING

TOMAS ANDRADE

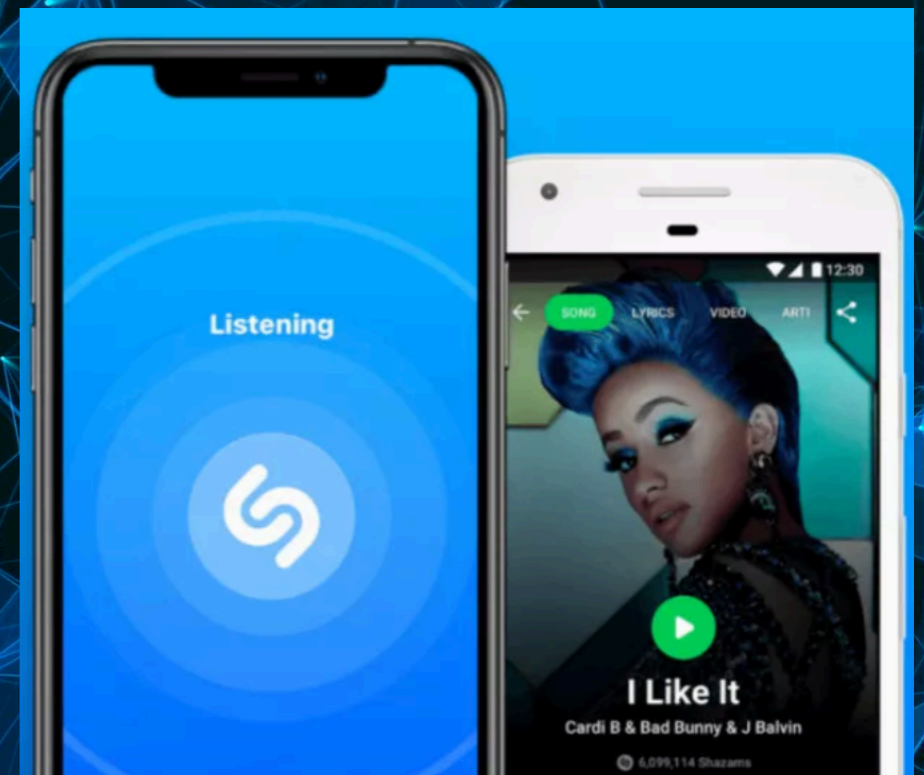
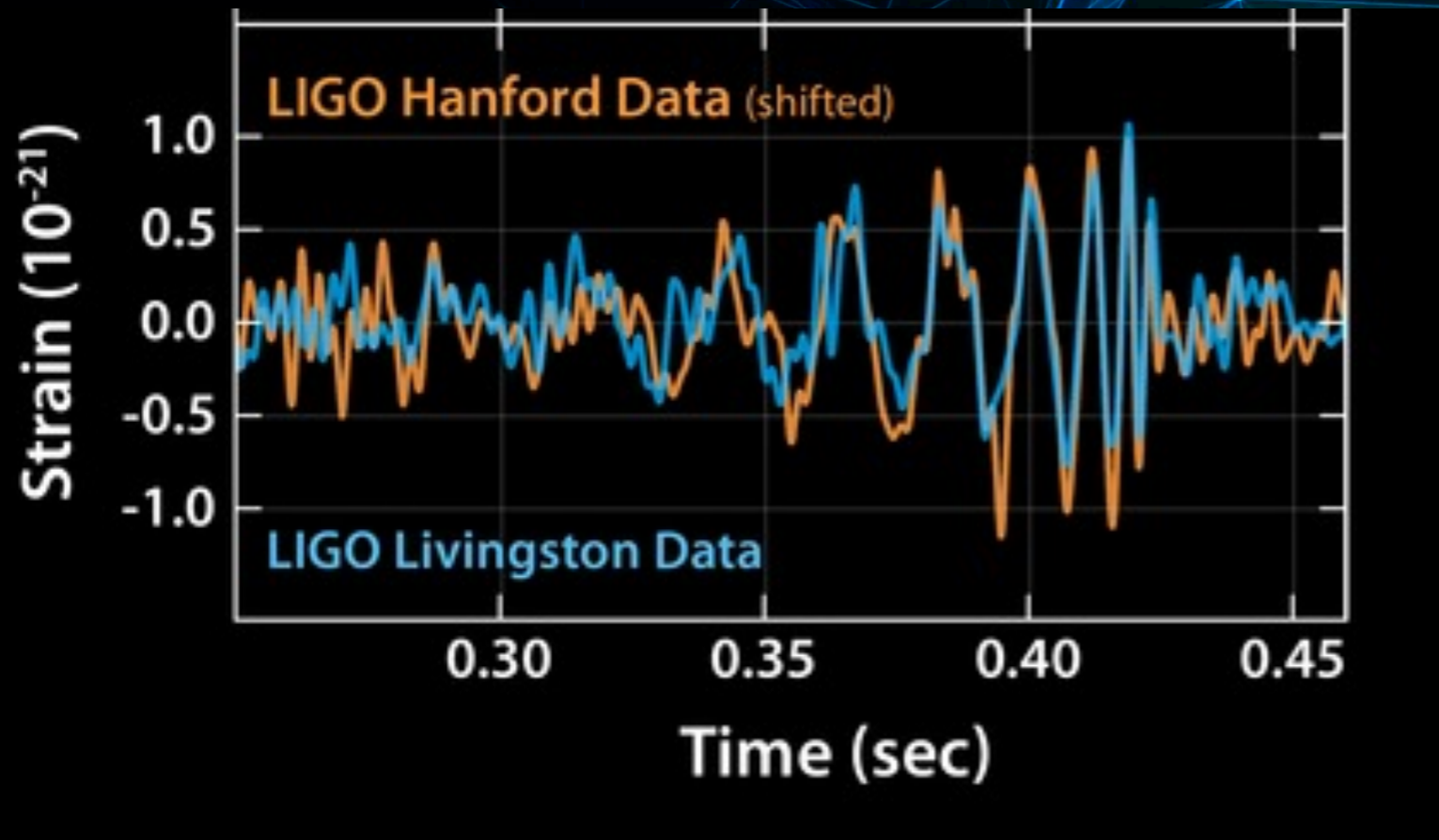


# All GW Talks Start Like This....

1.3 billion years ago in a galaxy  
far away ...

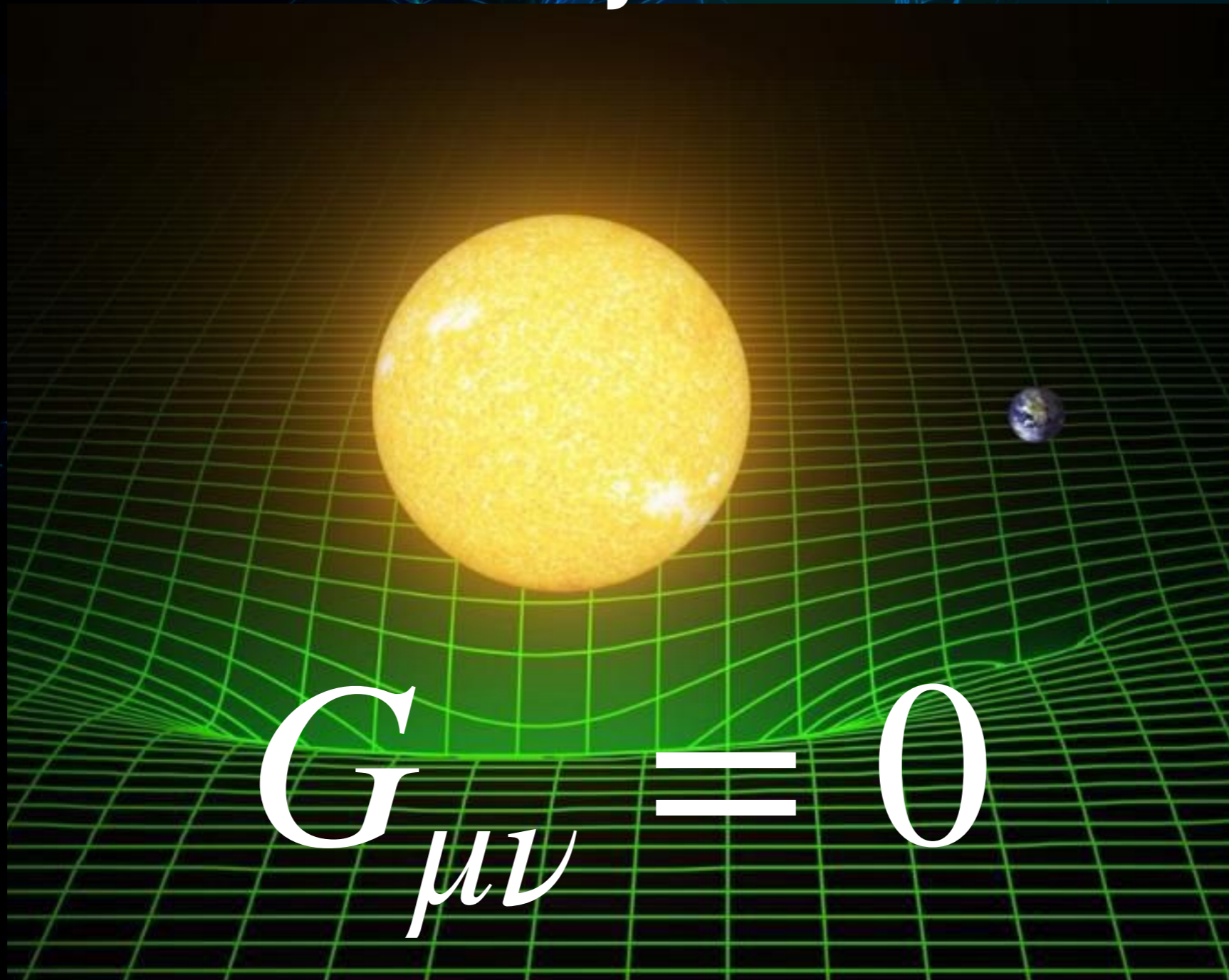


# What the Actual Data Looks Like



**Need a model!!**

We have a theory!



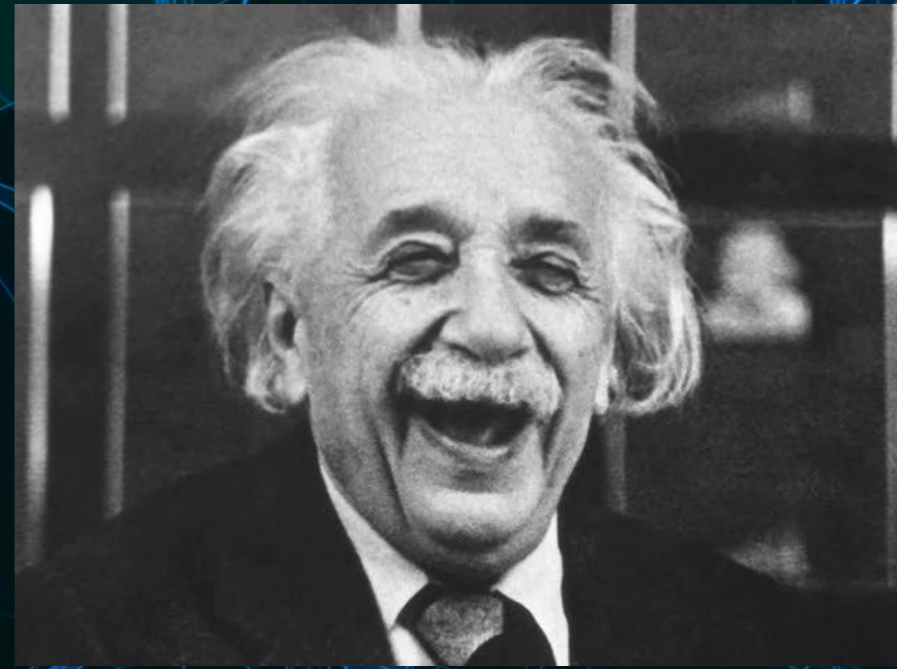
# Just do lots of simulations

## Expensive

days - weeks for BBH

Need about  $10^6$

10 parameters

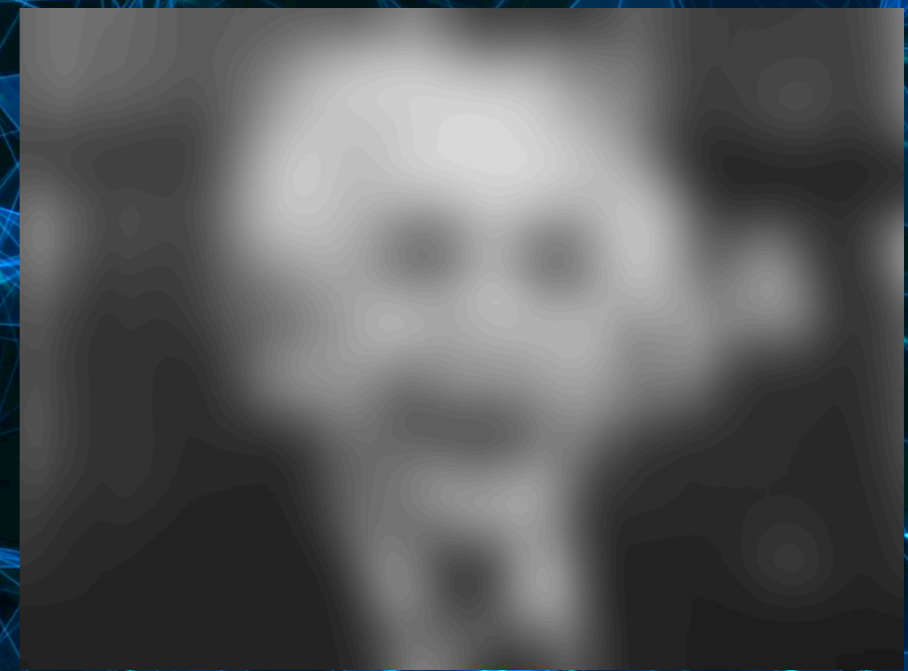


## Use approximants!

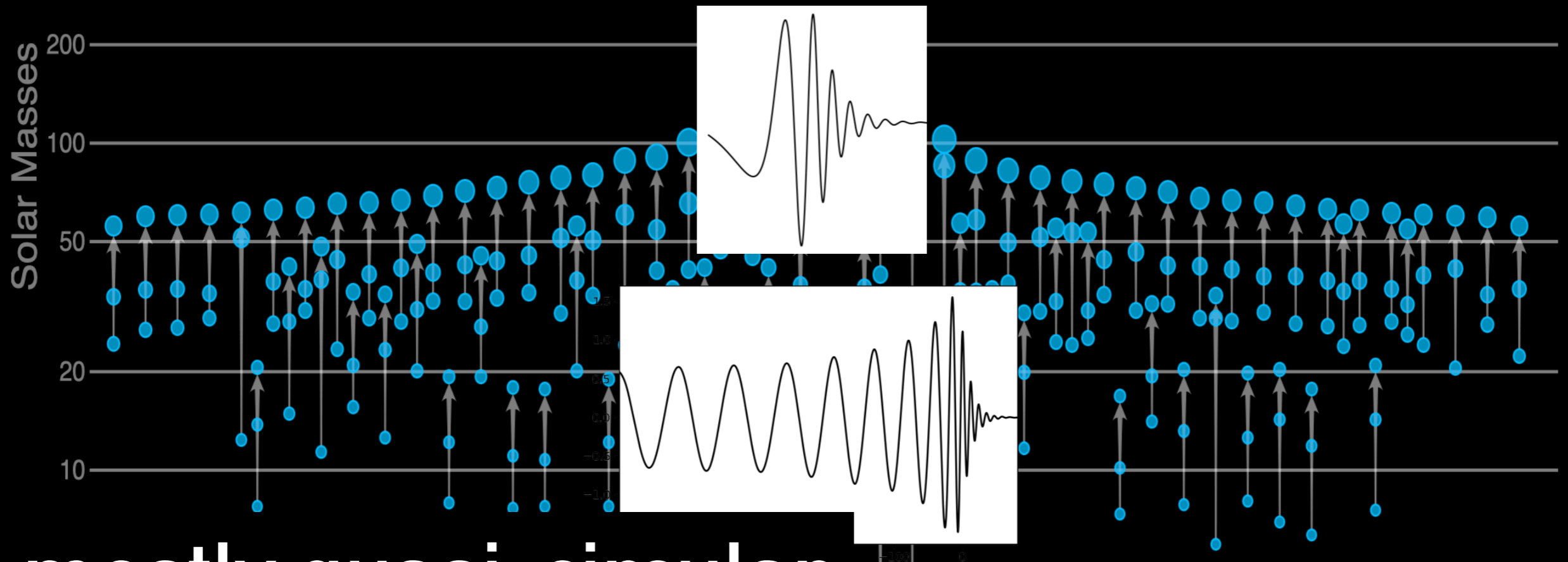
require calibration

**free parameters**

computation time  $\sim 1s$



# Masses in the Stellar Graveyard



mostly quasi-circular,  
but ...  
eccentricity  
precession  
high  $q$ , spins...

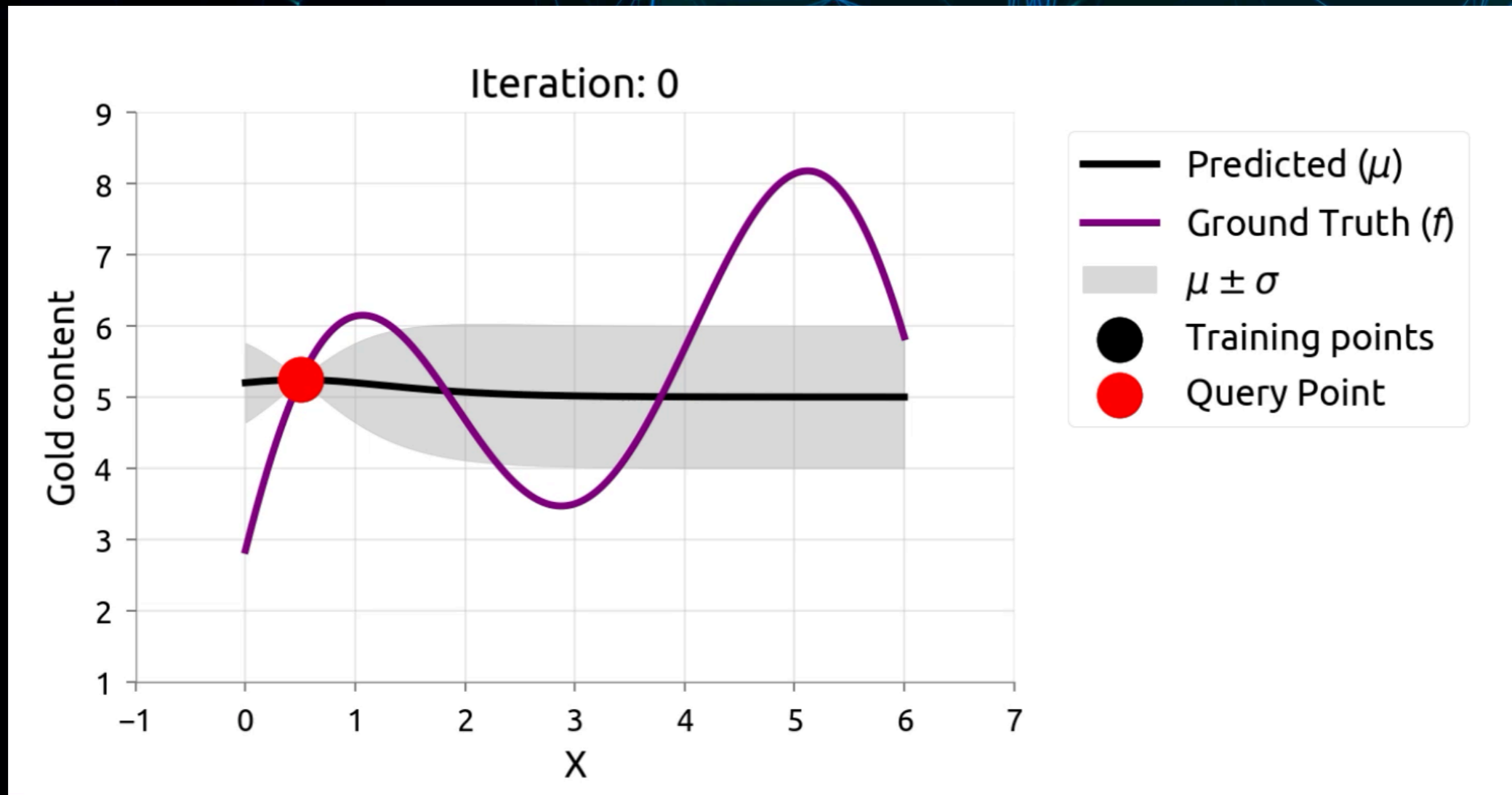
more diverse  
template banks



# EXPLORING PARAMETER SPACE

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# Active Learning with Gaussian Processes



Prediction + Uncertainty



# My job on one slide

$$h(X; t) \approx h_{approx}(c_i(X); t)$$

Construct training set  $\{X_A\}$  (domain knowledge)

Interpolate  $c_i(\{X_A\}) \rightarrow c_i(X)$  (ad-hoc fits)

**Use Active Learning  
+ Gaussian processes**



**HOW TO TEST THIS IDEA?**

# Proof of principle

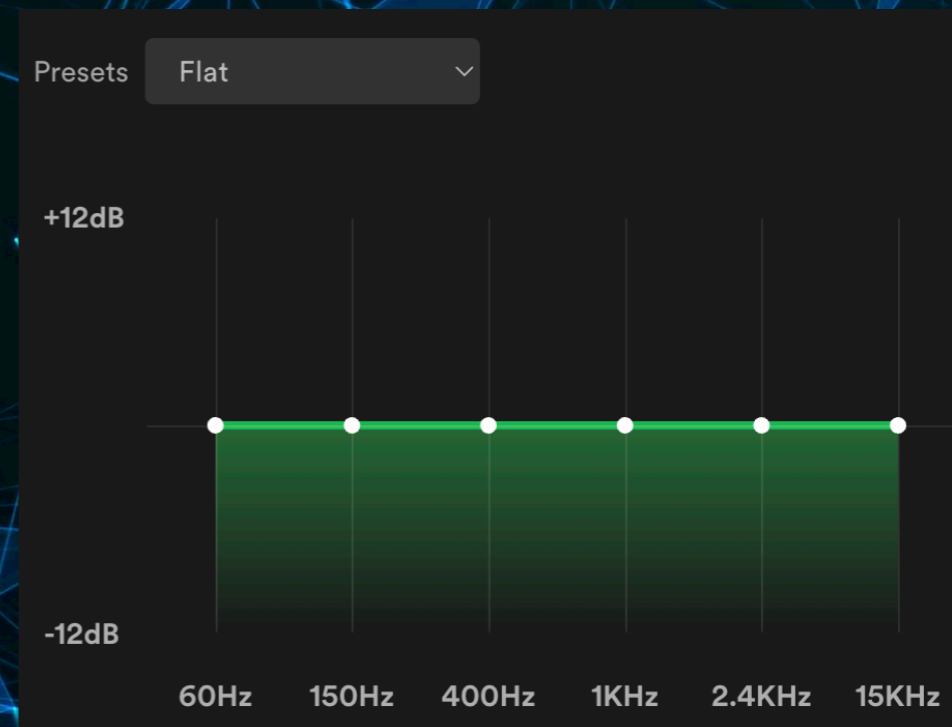
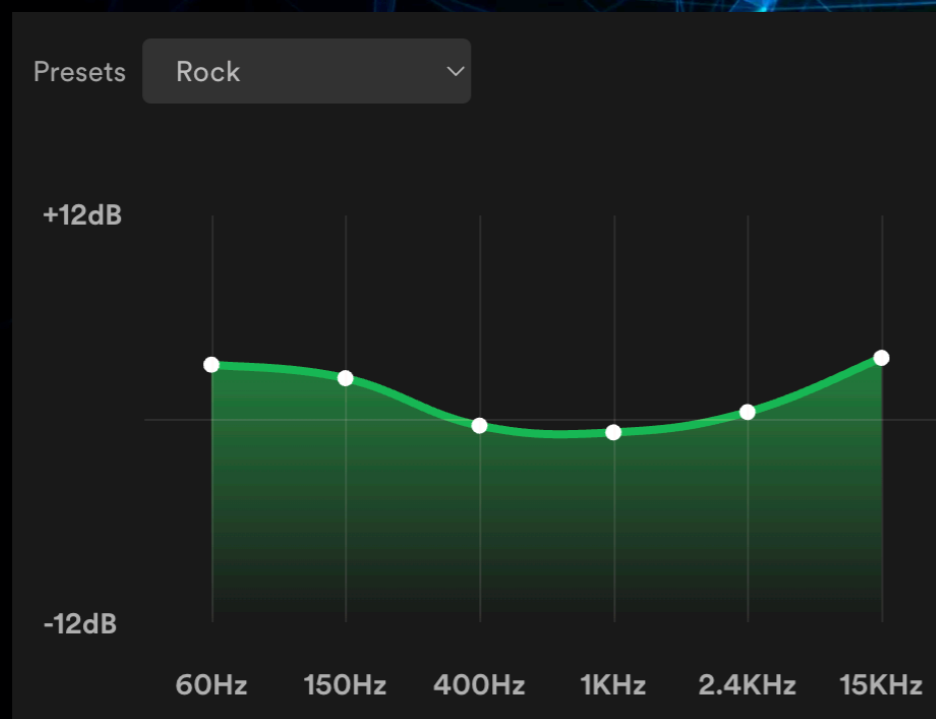
TA, R Gamba, J Trenado (2023)

Train an approximant with itself

quasi-circular

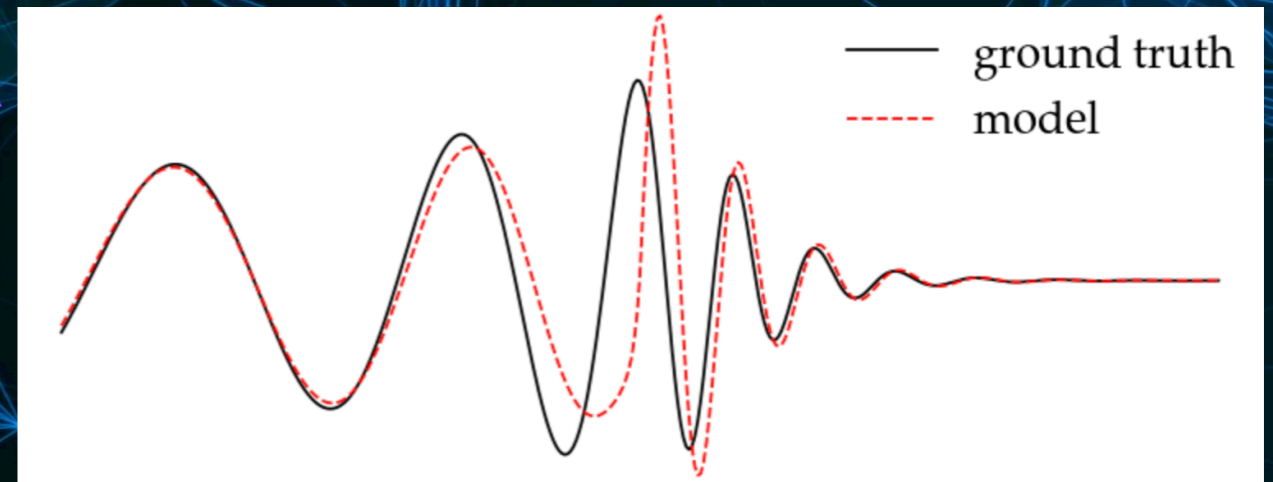
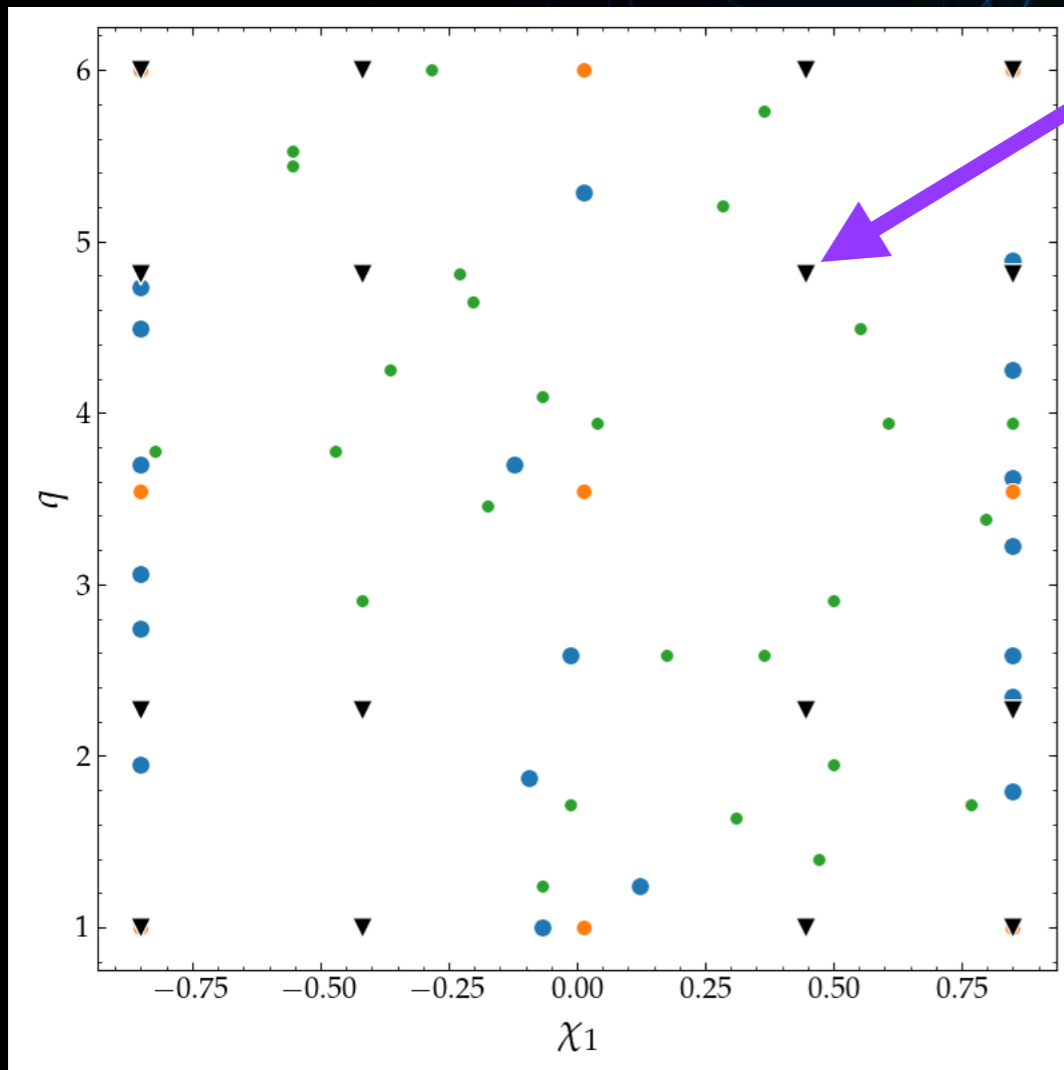
TEOBRResumS

$$X = (\chi_1, \chi_2, q)$$



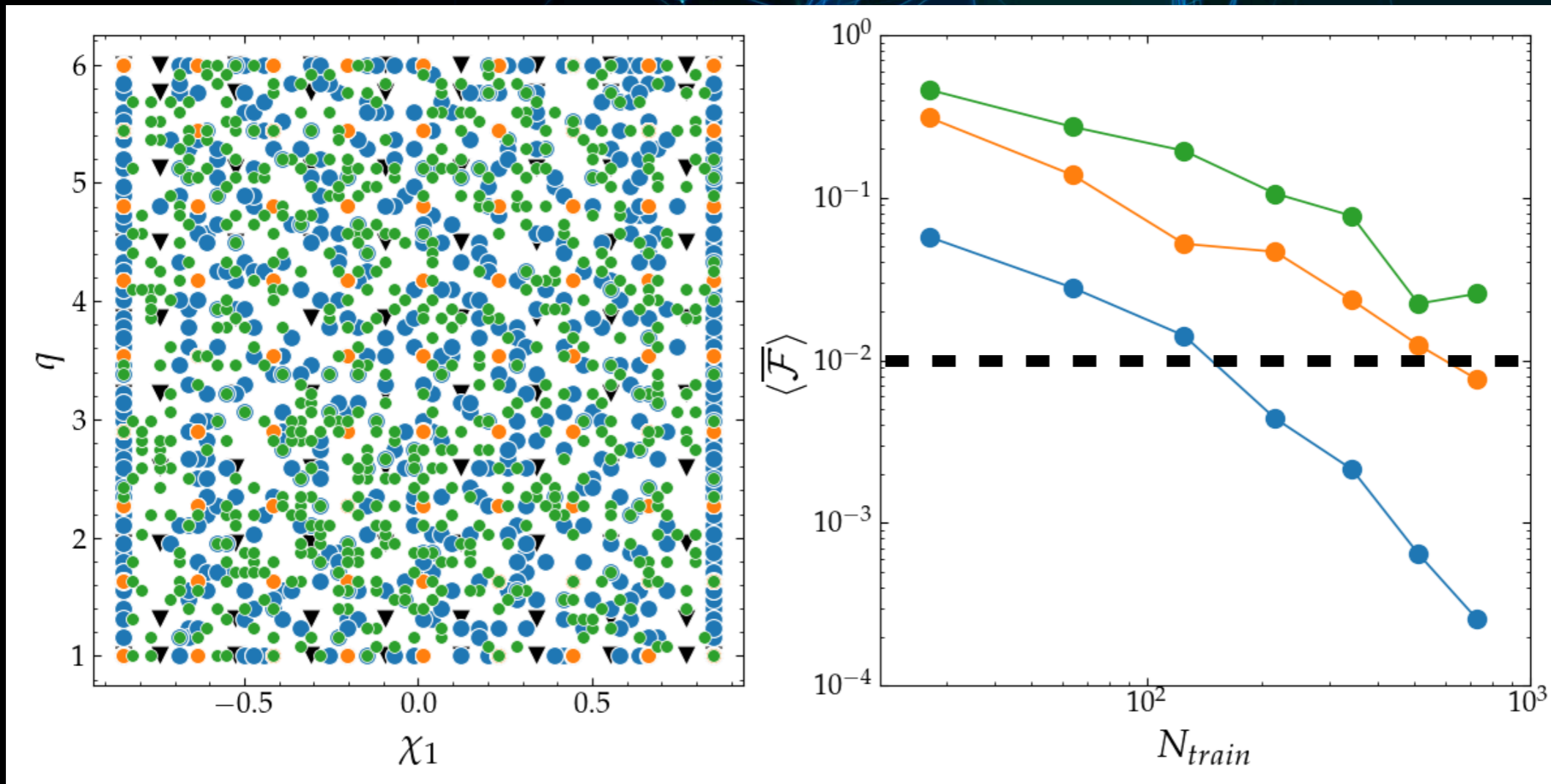
# Training strategies

● active    ● uniform    ● random    ▼ test



$$\overline{\mathcal{F}} = 0.12$$

# Evaluation

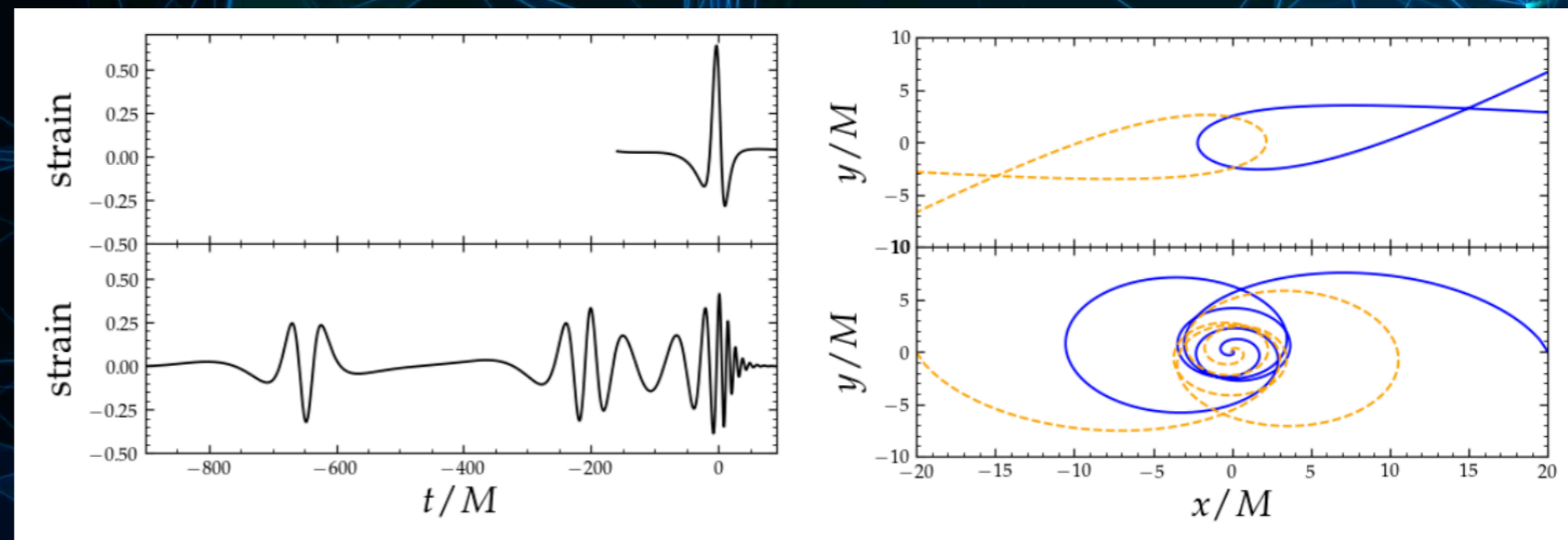


reduce cost by 4x

in parallel, reduce time by 3x

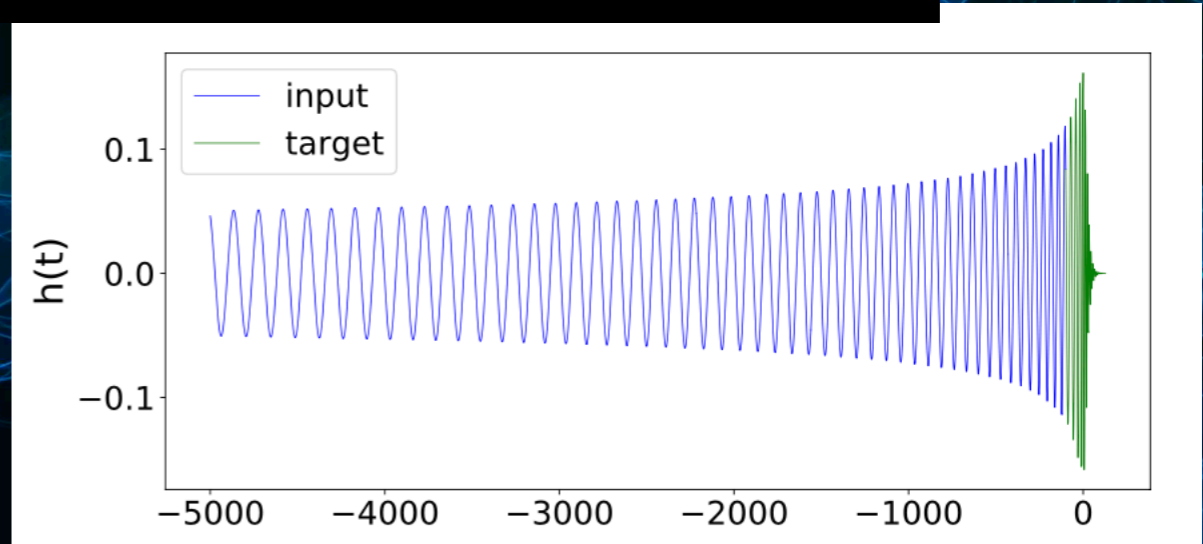
work in progress

# Surrogate models for dynamical captures



# Transformer based models for GW

(2021) Khan Huerta Zheng



**Take home**

**GW need modelling**

**NR informs approximants**

**Quasi-circular  $\rightarrow$  eccentric...**

**Need to explore parameter space**

**guide search with Active Learning**

Lee Cronin (2023)



You told me you  
worked on AI!!

“Autonomous  
Informatics”

**Thank you!**