

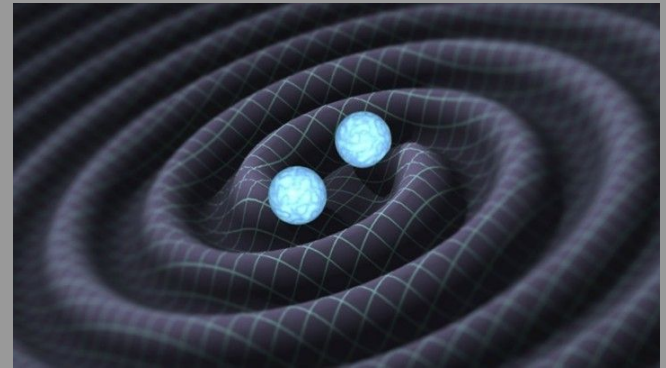
# Discovering the Underlying Distributions of Black Hole Populations

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Caltech / LIGO

# Overview

- Introduction to rates and populations
- Motivations
- Proposed mass distributions
- Recovering the mass distributions
- Conclusions / future work

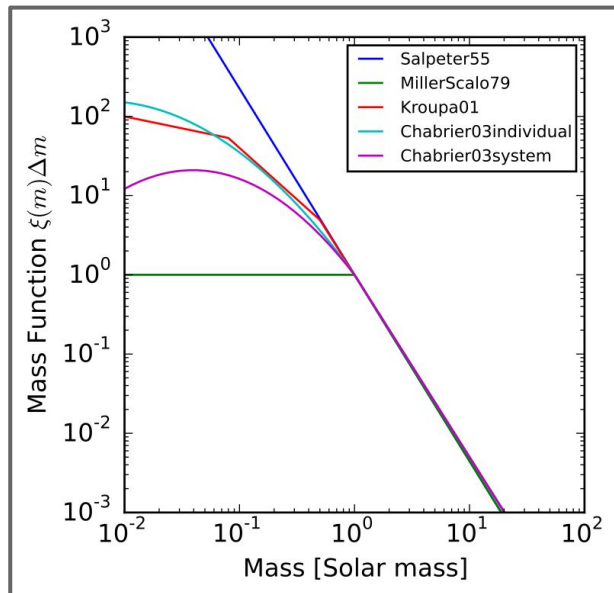
# Introduction to rates and populations

- Detector sensitivity  $\uparrow$ , distance we can hear GWs from compact binary mergers  $\uparrow$
- # of events detected will dramatically increase in the near future
- We expect tens, hundreds, or thousands of events
- Measure event rate density (in units of mergers/time/volume) as a function of mass, spin, and redshift (ignore spin for now)

$$\mathcal{R}(m_1, m_2, z) = \frac{dN}{dV_c dt_s}$$

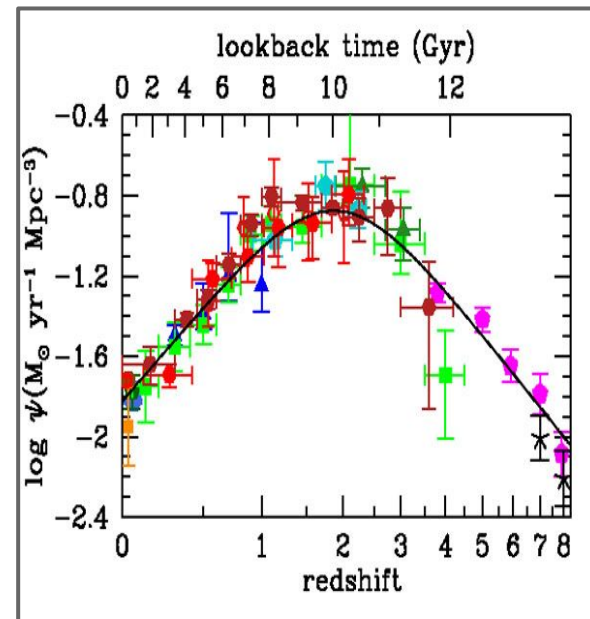
- Now is the optimal time to develop tools to use for this

# Models of $R(m_1, m_2, z)$ for Single Stars



Salpeter IMF, shown in **dark blue**, on a log scale. Describes the initial mass distribution for a stellar population. It appears linearly as its true nature is a power law.

(Johannes Buchner. "Initial Mass Function," Wikipedia.)



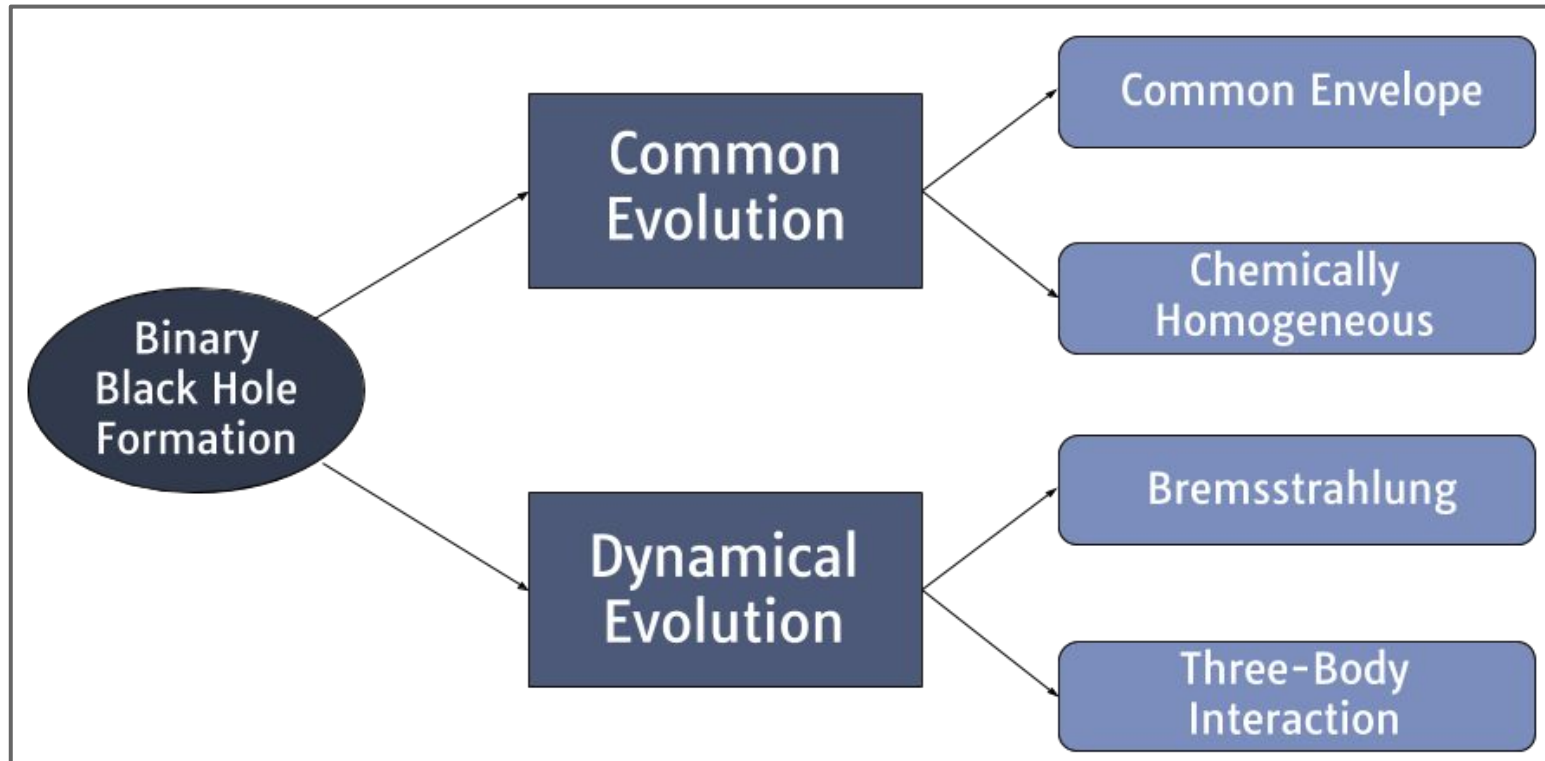
Madau-Dickinson star formation rate density (SFR/unit volume) as a function of redshift. The data points come from many other bodies of work. This distribution will be shifted left for BHs.

(Piero Madau, Mark Dickinson. *Cosmic Star Formation History*. 2014)

# Connecting Single Stars to BBHs

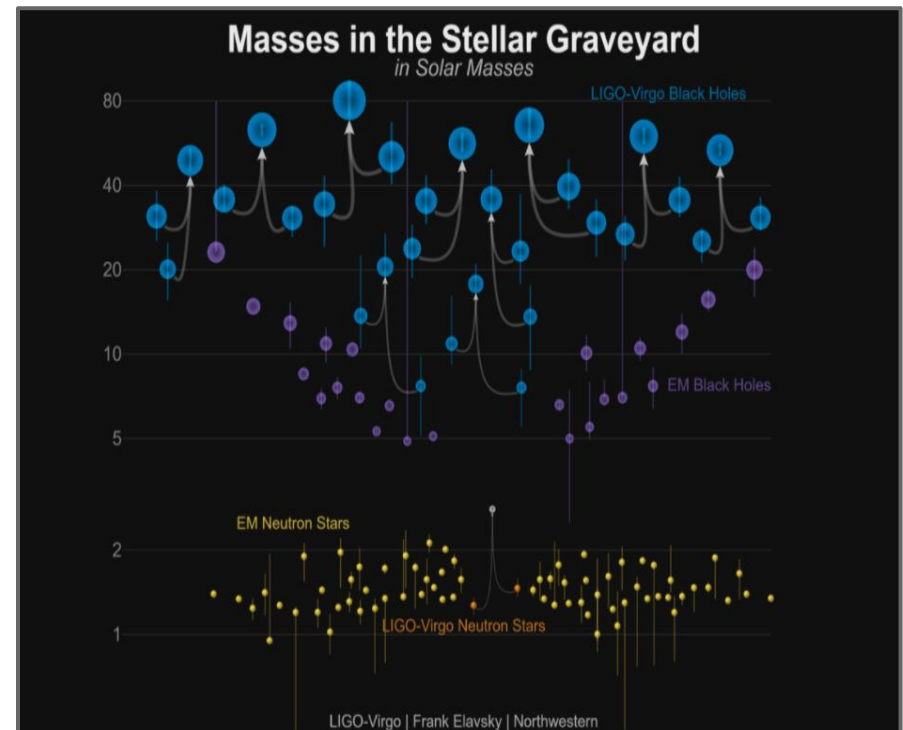


# Channels of BBH Formation



# Populations, Mass Distributions, and Mass Gaps

- $\sim 50\text{-}150M_{\odot}$ : Pulsational pair-instability supernovae (blows away significant portion of mass)
- $\sim 2\text{-}5M_{\odot}$ : Possible disparity between NS and BH masses
- $< 1M_{\odot}$ : Small likelihood that traditional stellar collapse would form BHs



# Mathematical Process

$$\frac{d\dot{N}(\lambda)}{d\vec{\theta}} = \mathcal{R}(1+z)^\gamma f(m_1|\alpha)f(m_2|\beta) \frac{dV_c}{dz} \frac{dt_d}{dt_s} \frac{1}{T_d} \quad (1)$$

Naturally-occurring event rate as a function of  $\theta = m_1, m_2, z,$   
 $\lambda = \alpha, \beta, \gamma$

$$\hat{N}_{true} = \int \frac{dN(\lambda)}{d\vec{\theta}} d\vec{\theta}, \quad (2)$$

Naturally-occurring, true number of events

$$\hat{N}_{det} = \int \frac{dN(\vec{\lambda})}{d\vec{\theta}} \mathcal{E}(\vec{\theta}) d\vec{\theta}, \quad (3)$$

Observed number of events

$$P(N|\hat{N}_{det}, \vec{\lambda}) = \frac{\hat{N}_{det}^N e^{-\hat{N}_{det}}}{N!} \quad (4)$$

Probability of true # of events given detected # of events



# Mathematical Process

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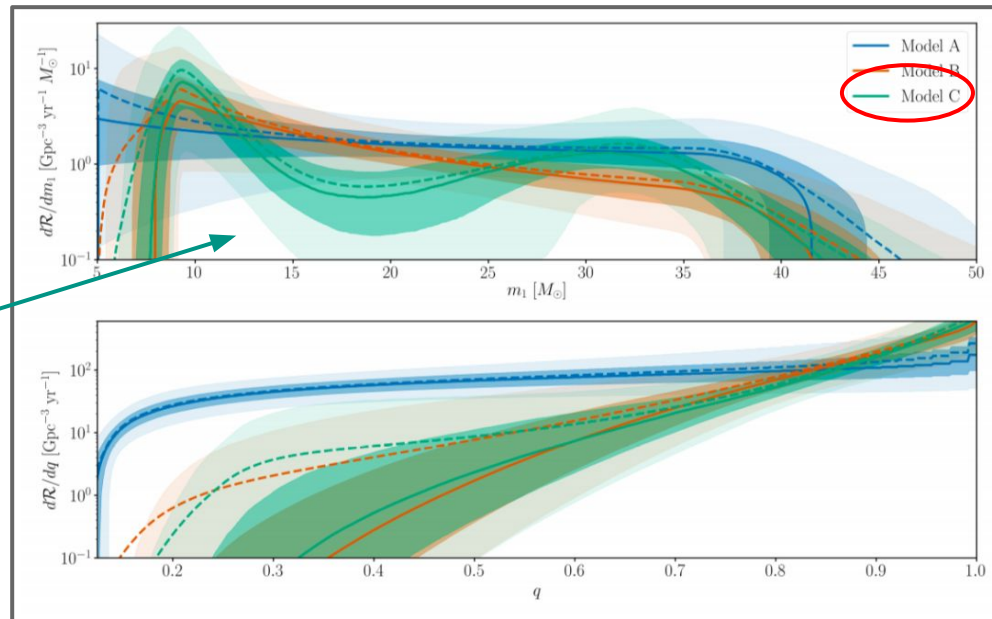
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Probability of true # of events given detected # of events

# Primary Mass Distribution



Power law component

Abbott, B. P., Abbott, R., Abbott, T. D., Abraham, S., Acernese, F., Ackley, K., ... & Agathos, M. (2019). Binary Black Hole Population Properties Inferred from the First and Second Observing Runs of Advanced LIGO and Advanced Virgo. *arXiv preprint arXiv:1811.12940*.

# Analysis Process

Generate optimal waveforms based on mass distributions in my model

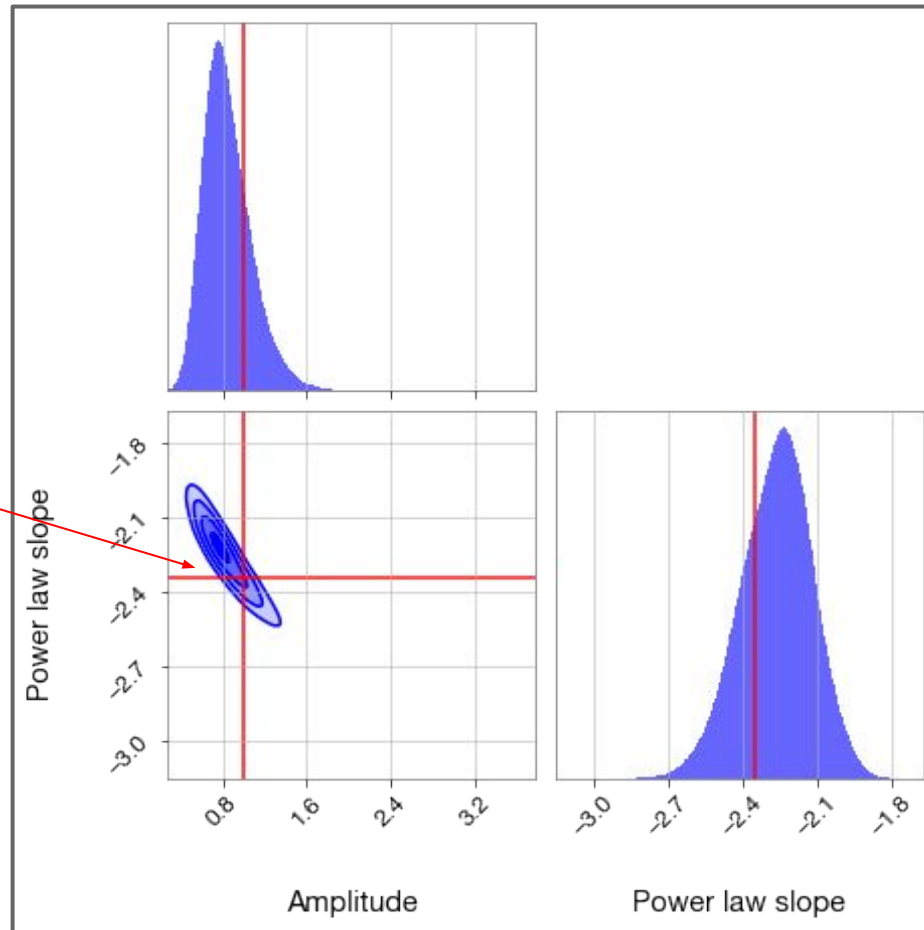
Retrieve the optimal SNR from each waveform, evaluate SNR threshold

Parameter estimation on the retrieved mass distributions to recover the hyperparameters  $\alpha, \beta, \gamma$

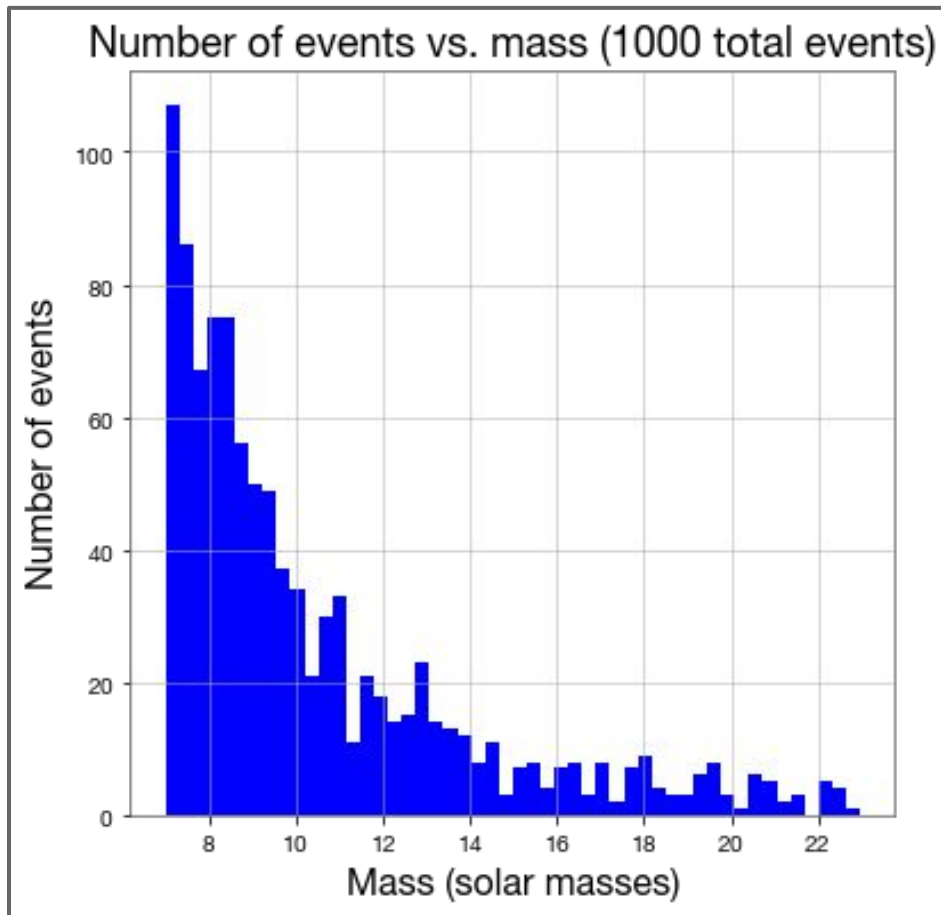
$$\rho^2_{opt} = \int \frac{\hat{h}^*(f)h(f)}{\mathcal{S}(f)} df$$

# Initial Parameter Estimation: Salpeter IMF

**Well-recovered!**



# Model C Power Law Component



**Parameter  
estimation  
to follow soon...**

# Conclusions / Future Work

- The known hyperparameters for the initial power law testing were well-recovered
- In the future, we will likely be able to conduct a similar parameter estimation on real populations of BBHs, and will be able to recover the underlying distribution
  
- Finishing Model C analysis / other more complex models
- Incorporating efficiency of detection
- Errors in the values of mass

# Acknowledgements

- Alan Weinstein
- Jonah Kanner
- Liting Xiao
- Tom Callister
- Shreya Anand
- Ryan Magee



# Other Slides

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# Bayesian Inference

$$p(\theta|d) = \frac{\mathcal{L}(d|\theta)\pi(\theta)}{\mathcal{Z}}$$



Bayes' Theorem

$$\mathcal{Z} \equiv \int d\theta \mathcal{L}(d|\theta)\pi(\theta)$$



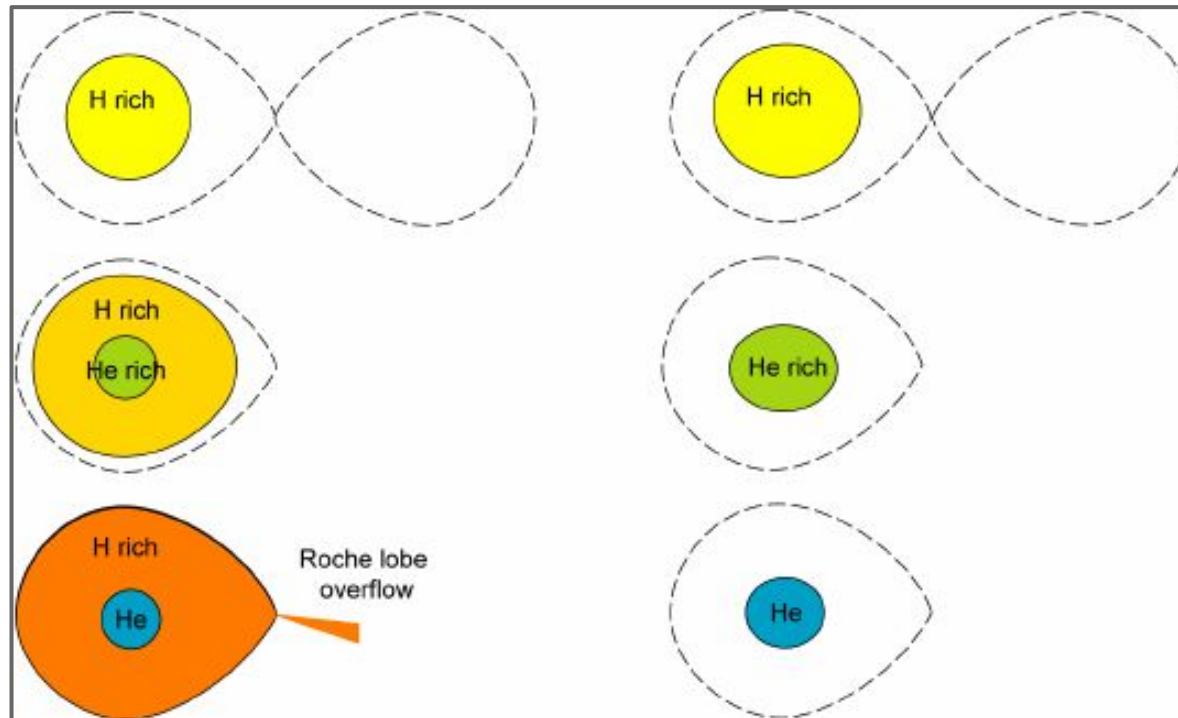
Evidence

$$BF_2^1 = \frac{\mathcal{Z}_1}{\mathcal{Z}_2}$$



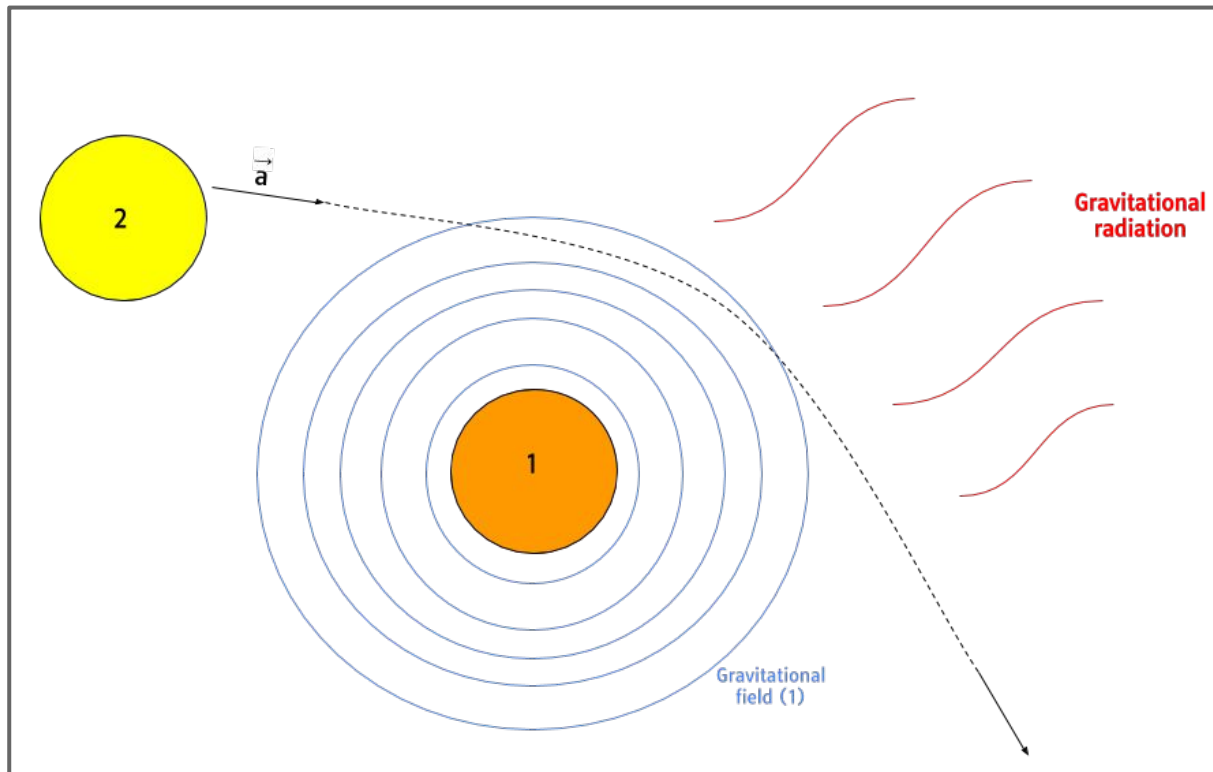
Bayes Factor

# Common Envelope vs. Chemically Homogeneous

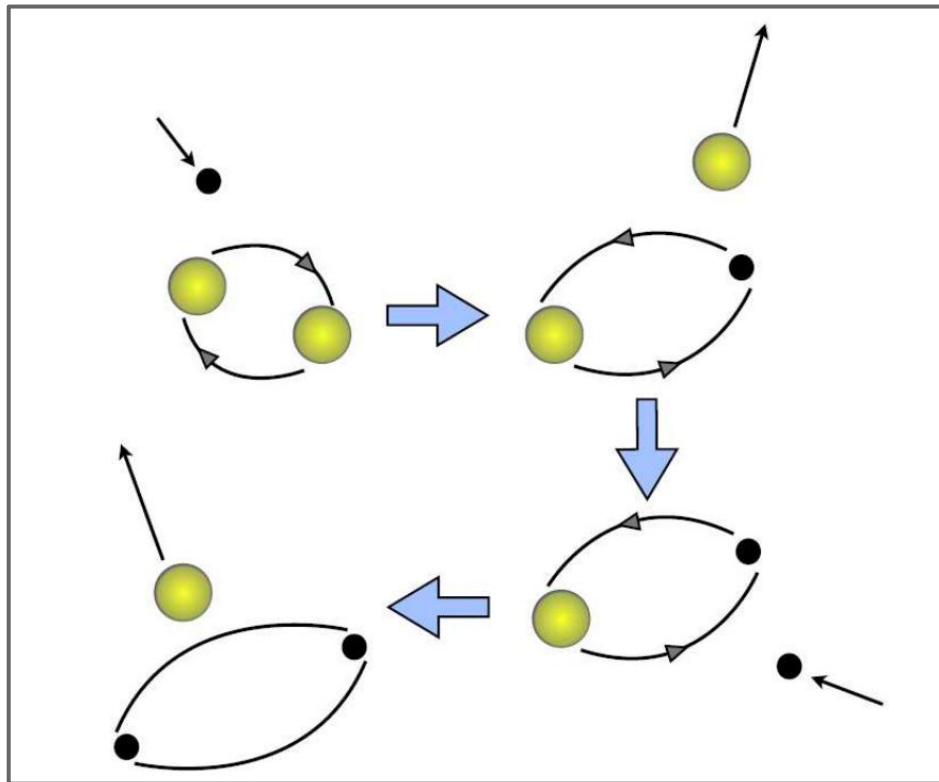


(de Mink, 2008)

# Bremsstrahlung



# Three-Body Interaction



(Banerjee, 2016)