Binary Populations







Binaries and Populations II



Previously

- What is stellar population synthesis?
- Why is it useful?
- The big binary parameter space
- Sampling the parameter space
- Initial distributions in single and binary stars
- Fast synthetic stellar evolution models

What's coming next

- Observations: the things to compare with
- Selection effects
- How to compare models to obs.
- The power of population synthesis: binary stars
- My binary_c code → examples for you!
- Population synthesis case studies

Observations

- However good our theories, astrophysics is still very much observationally driven
- These are usually either surveys e.g.
 - Hipparcos, Gaia
 - SDSS, APOGEE, Galah
 - COROT, Kepler
- Or "random" pointings toward objects of interest (often called "surveys"!)
- Always SELECTION EFFECTS



 Volume limit is simplest, you need do nothing if you think the sample has enough objects

• Magnitude limits you can model with a cut off or

completeness function

(cf. Malmquist bias)

• A complete survey is rare

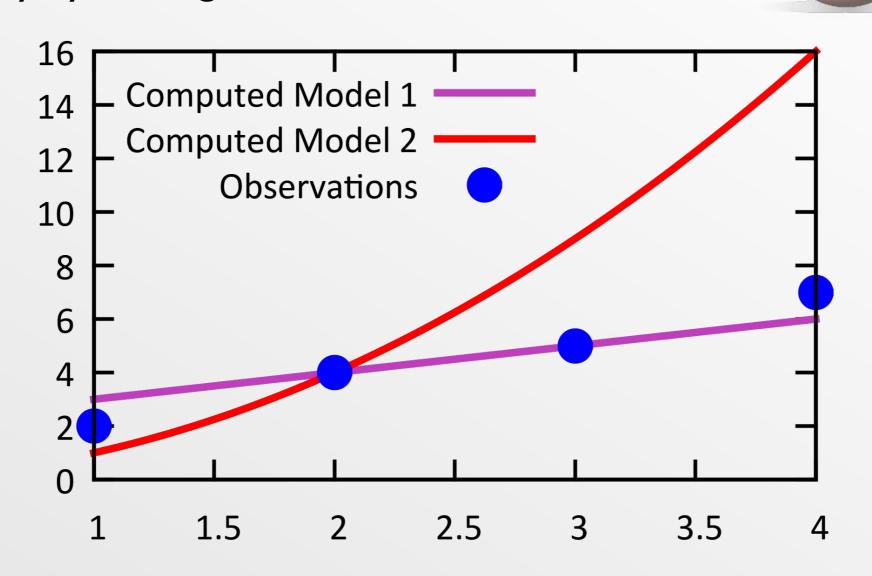
but possible, e.g. strong

barium stars (more on these later!)

• Random choice may be ok if sufficient number of stars to justify the risk. This may also be all you can get.

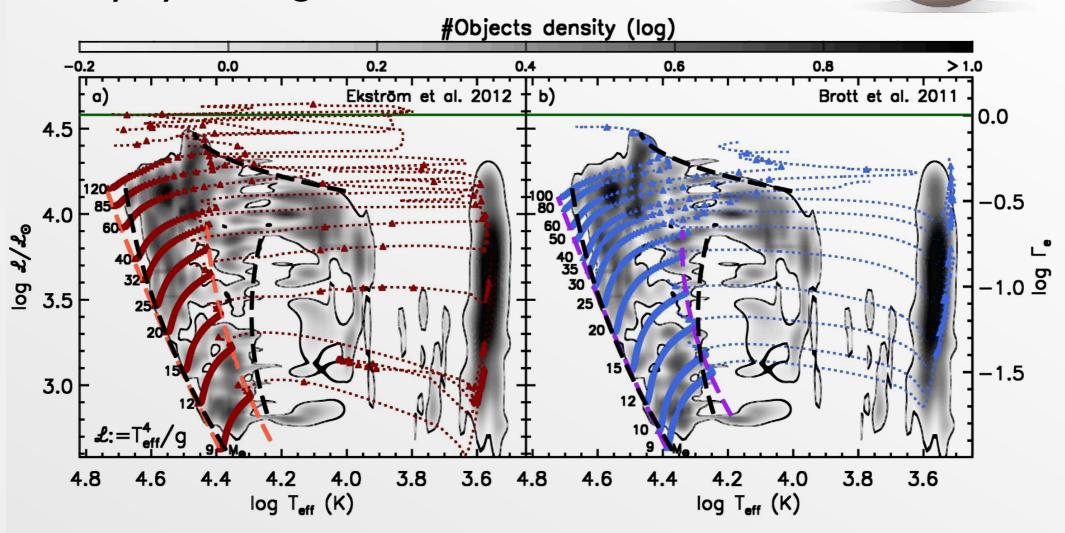
Models vs observations

• By eye fitting: which is the best match?



Models vs observations

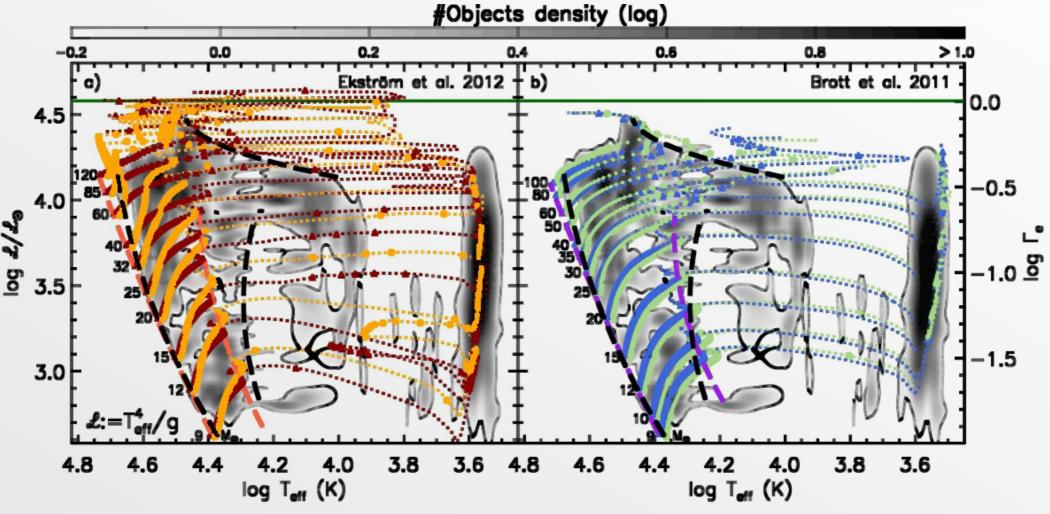
By eye fitting: which is the best match?



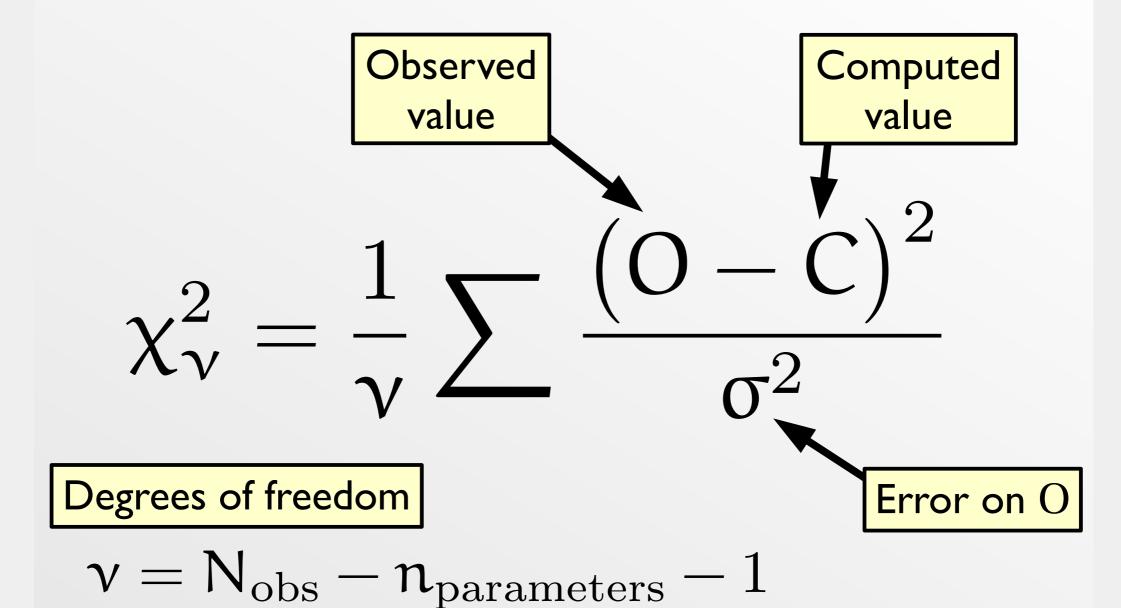
Models vs observations

• By eye fitting: Rotating models are better?

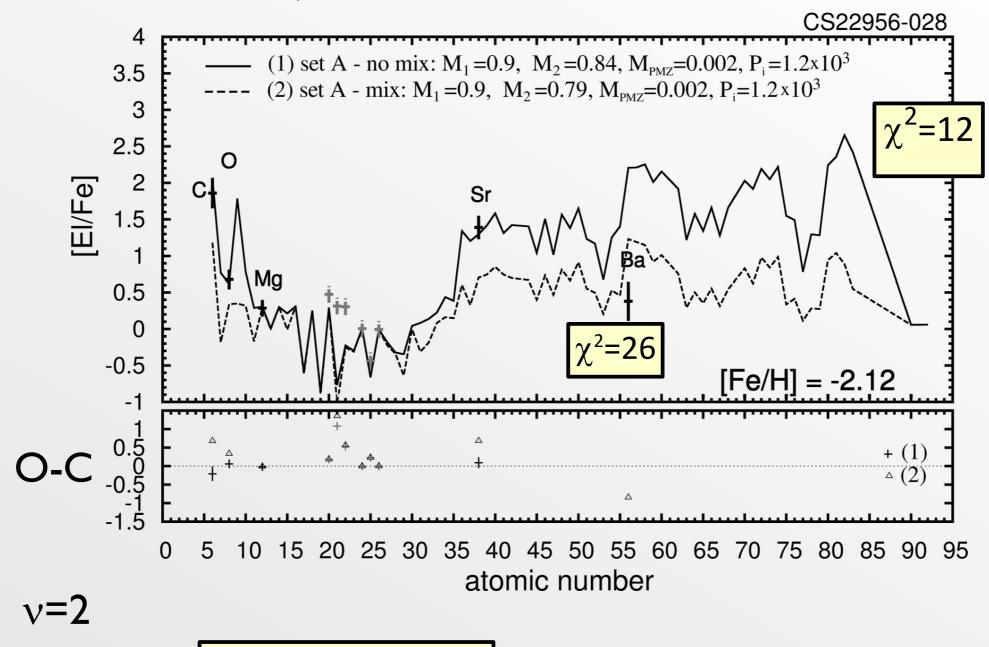




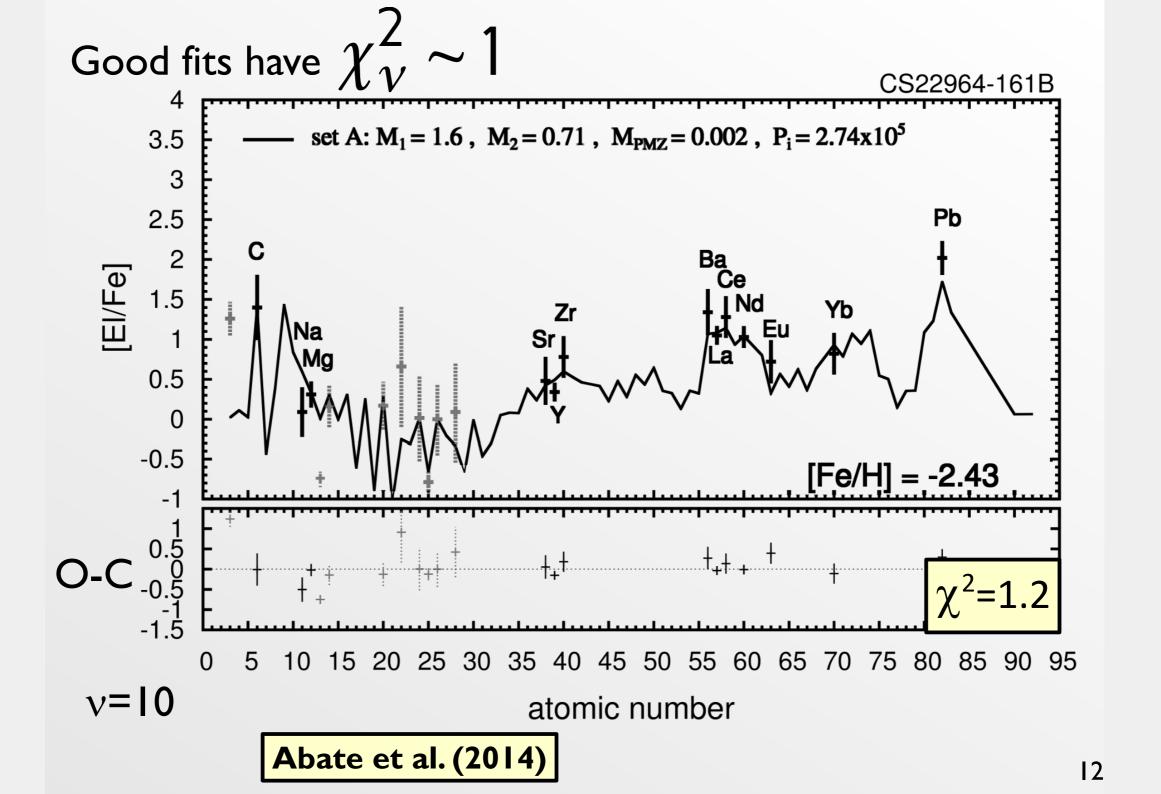
Models vs observations: χ² tests



Bad fits have $\chi_{\nu}^2 \gg 1$



Abate et al. (2014)



Amoeba/Downhill Simplex $(x-7)^2+(y+3)^2$ Minimise function O-C

X

Other matching algorithms

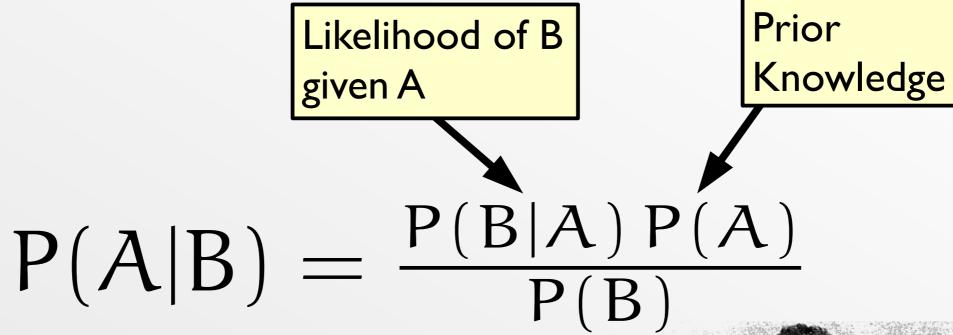
- Genetic
 - Minimize fitness functions
- Markov-Chain Monte Carlo
 - → used often e.g. in cosmology
- Etc. → whole branch of numerical analysis
- Latest applications to (many) stars:

Bayesian methods



Bayesian Methods

• Is O-C good enough? Bayes says no.



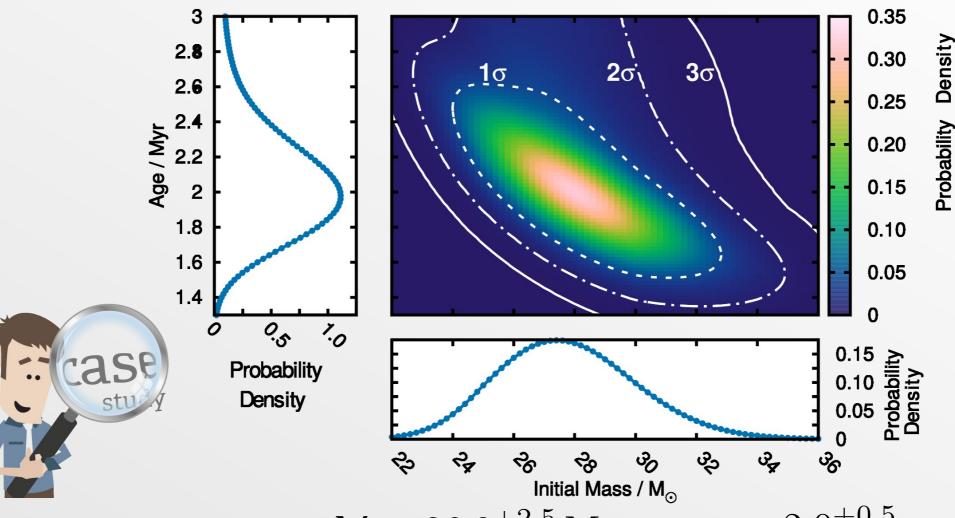
Can be used to compare O and C and provide a *distribution* of the (many) best fitting parameters



$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

Case study: Given logL, T_{eff} and logg, of star V3903 Sgr A what is M?

Torres et al. (2010): $M = 27.27 \pm 0.55 \,\mathrm{M}_{\odot}$

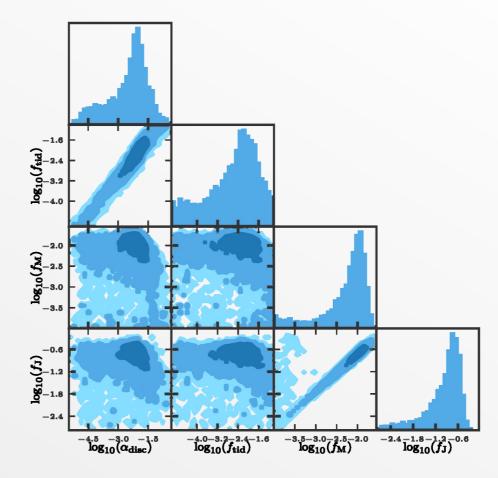


Best model fit:
$$M = 26.8^{+2.5}_{-2.0} \,\mathrm{M}_{\odot}$$
 age $= 2.0^{+0.5}_{-0.4}$

$$age = 2.0^{+0.5}_{-0.4}$$

Bayes + (MC)MC is easy these days

- Python packages abound
- emcee is popular
 - → there are many others!
- better comparison O-C
 - → error bars
 - → bootstrap
 - → constraints

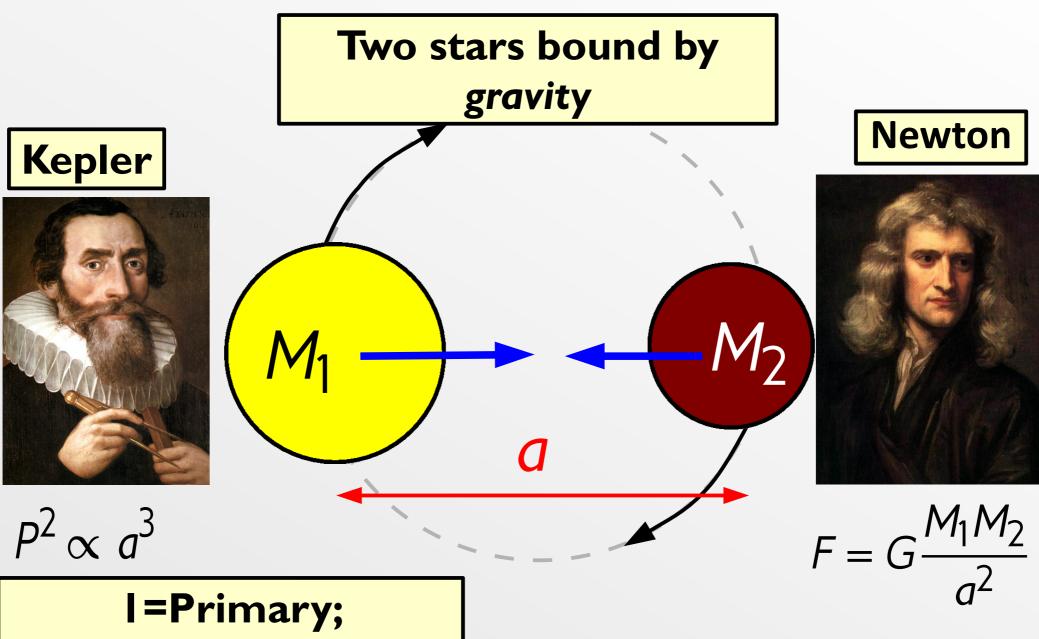


binary_c-python +
emcee + pygtc

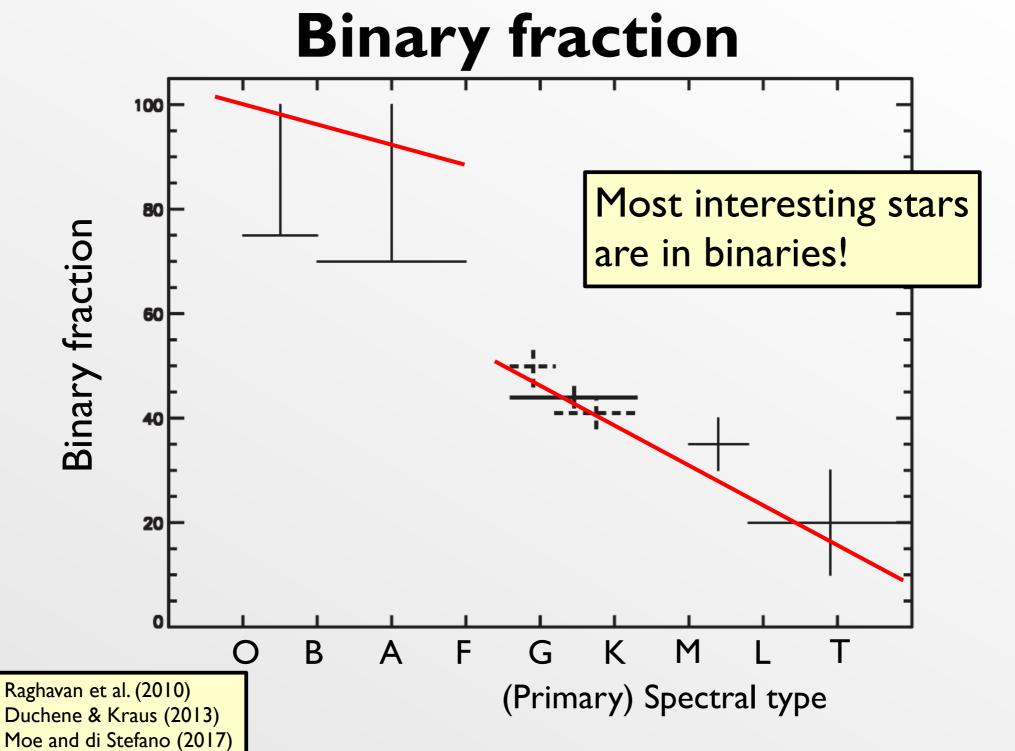
Recap: We have discussed

- What a population of stars is
- How to distribute stars (initial mass function etc.)
- How to make population statistics
- How to compare the population with observations
- But we haven't discussed a powerful aspect of population synthesis...

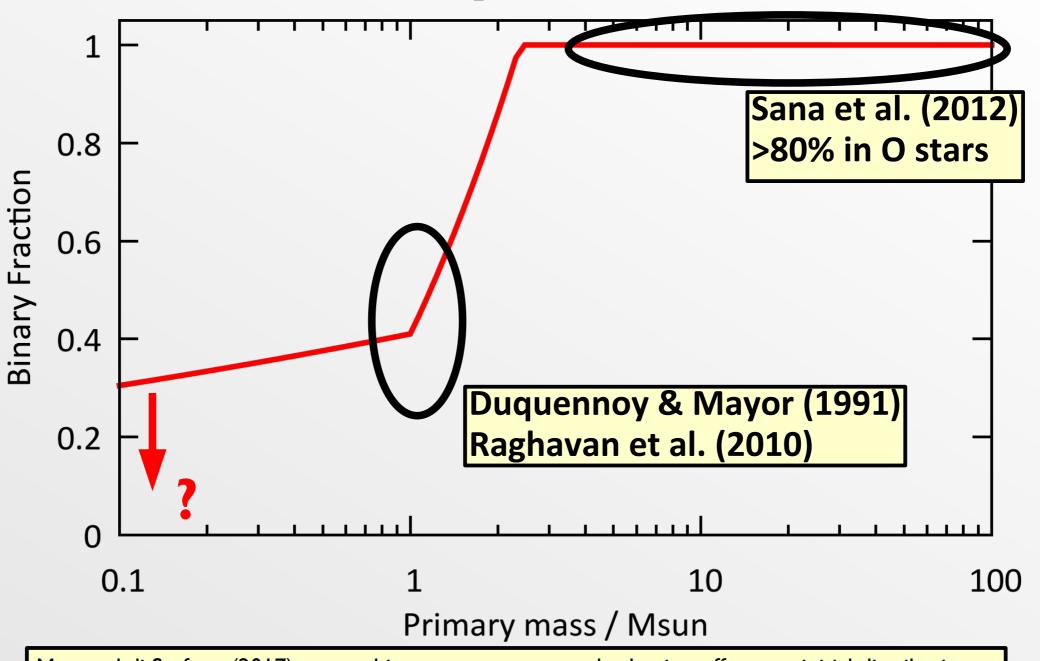
Binary stars



2=Secondary



Binary fraction



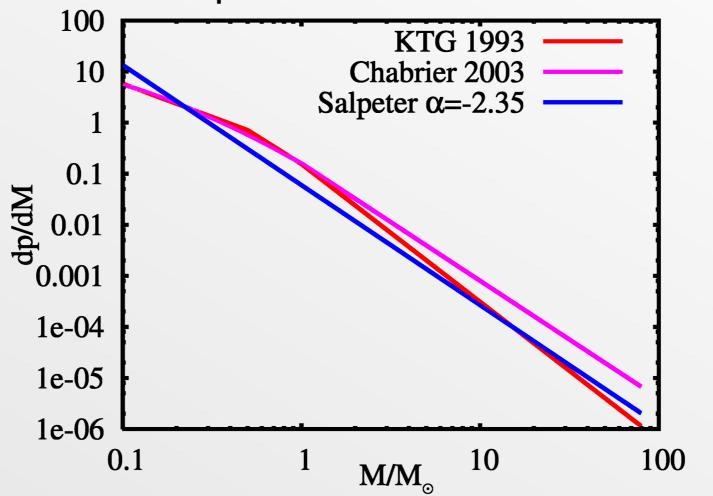
Moe and di Stefano (2017) → combines many surveys and selection effects → initial distributions

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Binary star parameter space

$$P(M_1, M_2, a) = \psi(M_1) \phi(M_2) \chi(a)$$

• Primary mass M_1 ~ follows single-star IMF $\psi(M_1)$

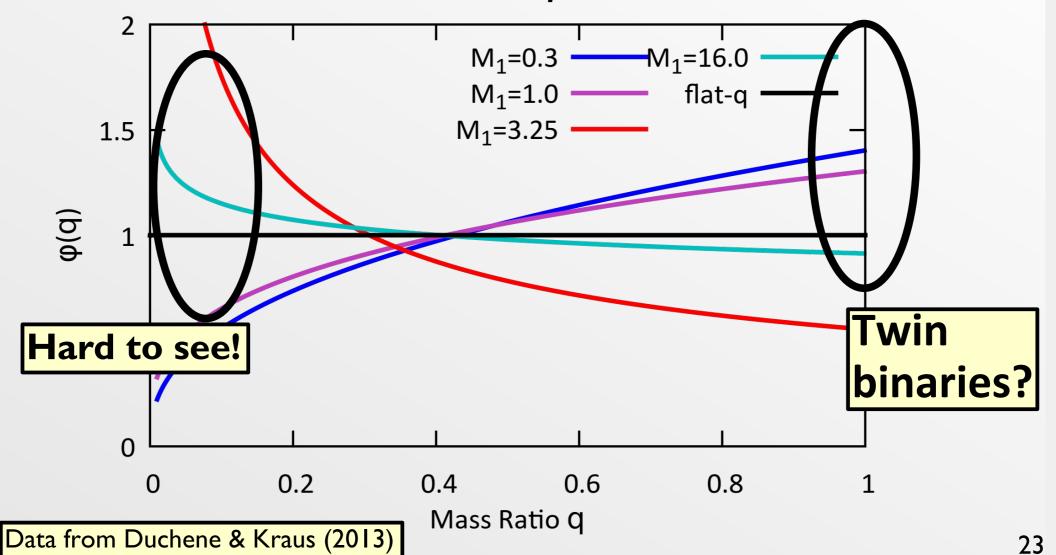


Universal? Probably not ... e.g. GallMF module

Binary star parameter space

$$P(M_1, M_2, a) = \psi(M_1)\phi(M_2)\chi(a)$$

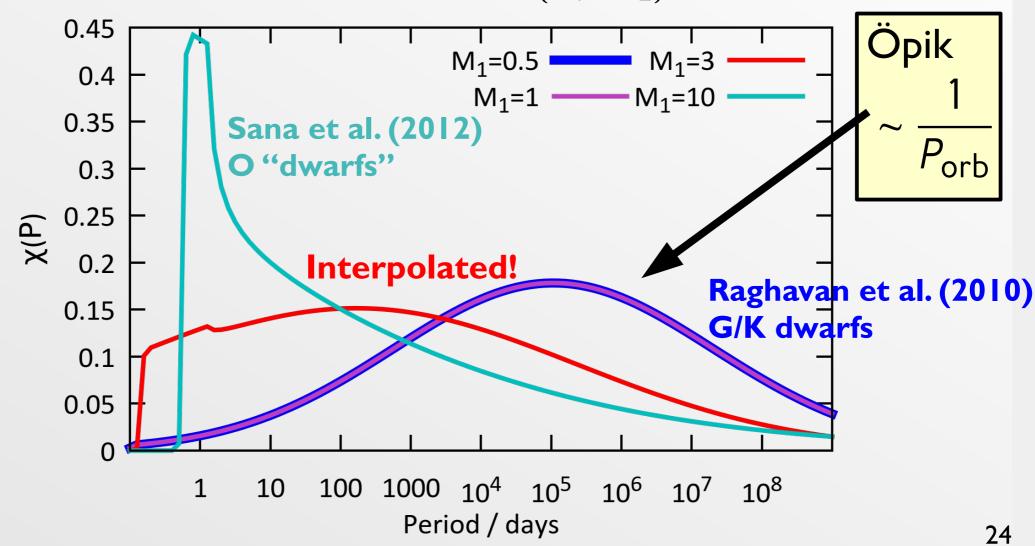
• Secondary mass $M_2 = q \times M_1$



Binary star parameter space

$$P(M_1, M_2, a) = \psi(M_1) \phi(M_2) \chi(a)$$

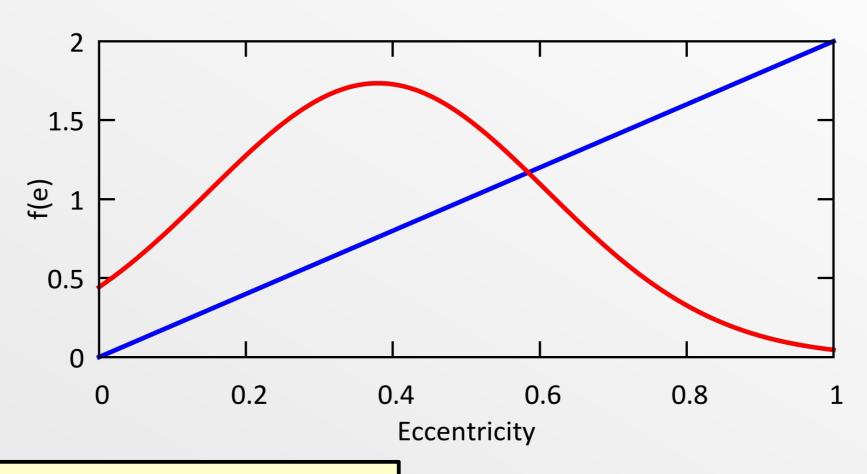
• Separation/Period
$$P_{\text{orb}}^2 = \frac{4\pi^2}{G(M_1 + M_2)}a^3$$



Binary star parameter space $P(M_1, M_2, a, e) = \psi(M_1) \phi(M_2) \chi(a) \times f(e)$

$$P(M_1, M_2, a, e) = \psi(M_1) \phi(M_2) \chi(a) \times f(e)$$

Eccentricity



Safe(ish) to assume e=0 ...

Binary statistics reminder: not trivial!

Number of stars of interest:

$$N = f_{\text{bin}} \sum_{M_1} \sum_{M_2} \sum_{a} \sum_{e} \sum_{t} S(t) \psi \left(M_1 \right) \phi \left(M_2 \right) \chi (a) f(e) \bar{\delta}(t) \delta t + \left(1 - f_{\text{bin}} \right) \sum_{M} \sum_{t} S(t) \psi \left(M \right) \bar{\delta}(t) \delta t$$

Usually: set e=0, S=1, fix $Z\rightarrow\delta(t)$ and calculate ratios

(see previous lecture)

But still a lot of work: at least a quadruple sum.

Now you see why fast synthetic codes are useful!

Binaries: fundamental properties

Masses: M_1 M_2

Energy: $E = \frac{1}{2}M_1|\dot{\mathbf{r}}_1|^2 + \frac{1}{2}M_2|\dot{\mathbf{r}}_2|^2 - \frac{GM_1M_2}{r}$

Angular momentum:

$$\mathbf{J} = M_1 \mathbf{r}_1 \times \dot{\mathbf{r}}_1 + M_2 \mathbf{r}_2 \times \dot{\mathbf{r}}_2$$



Tides: synchronization and circularization

• Torque
$$\Gamma = -\frac{\Omega - \omega}{\tau_{\text{diss}}} q^2 M R^2 \left(\frac{R}{a}\right)^6$$

• Hence lowest energy state: $\Omega = \omega$ and e = 0

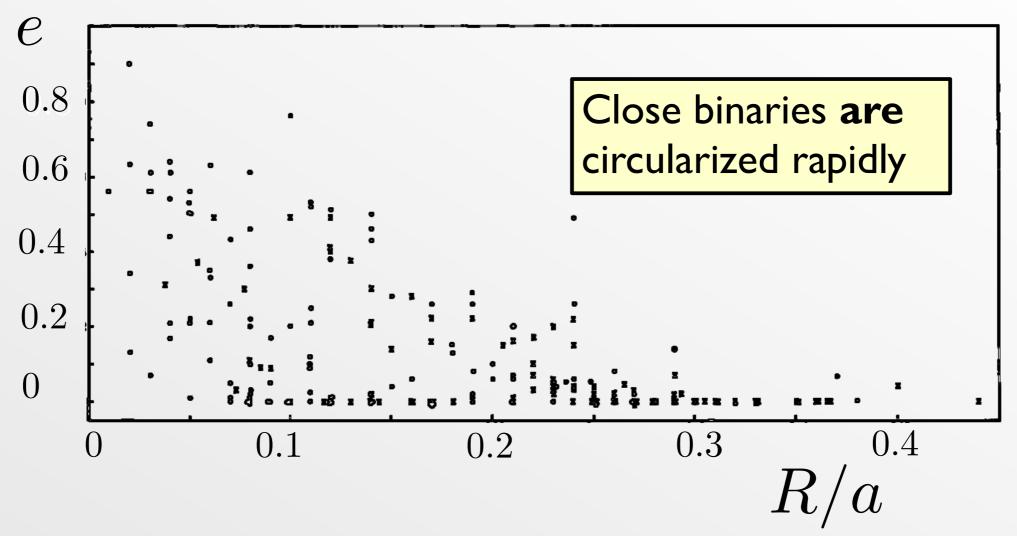
$$\tau_{\text{sync}}^{-1} = \frac{-\Gamma}{I\Omega} = \frac{1}{\tau_{\text{diss}}} \frac{\Omega - \omega}{\Omega} q^2 \frac{MR^2}{I} \left(\frac{R}{a}\right)^6$$

$$\tau_{\text{circ}}^{-1} = \frac{\dot{e}}{e} = \frac{1}{\tau_{\text{diss}}} \left(9 - \frac{11}{2} \frac{\Omega}{\omega} \right) q(1+q) \left(\frac{R}{a} \right)^{8}$$

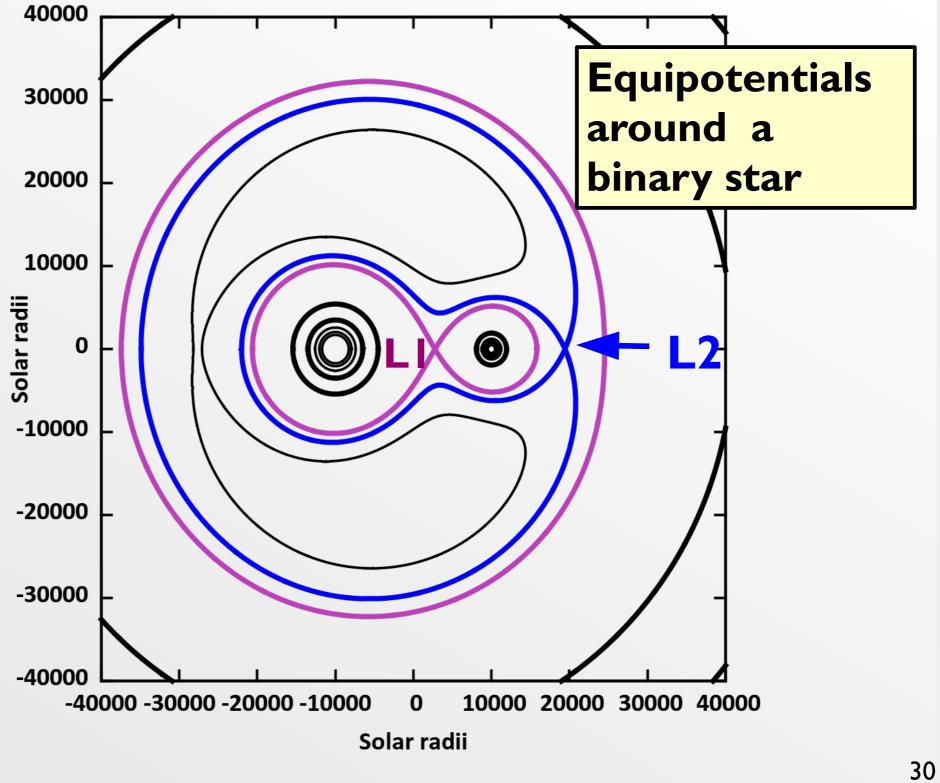
e.g. Zahn 1972, 1980s Hut 1981, book of Tassoul etc.

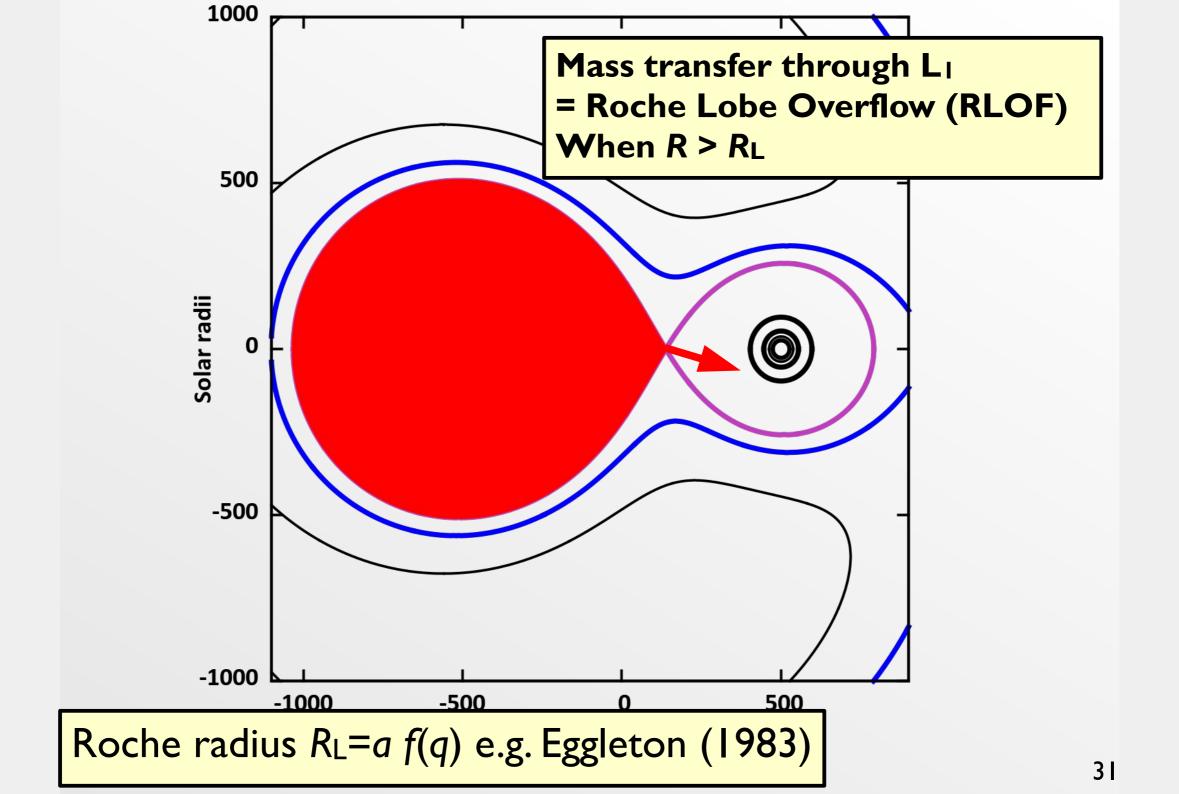


e vs R/a (unevolved binaries)

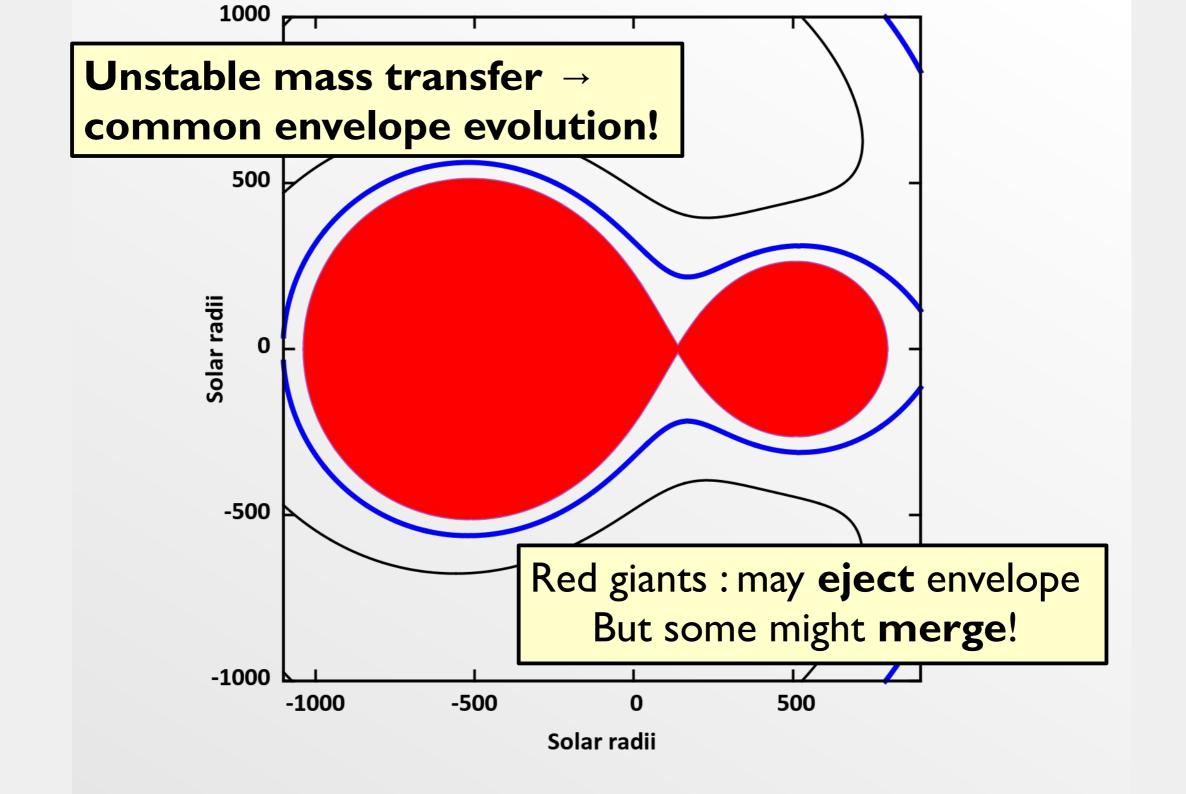


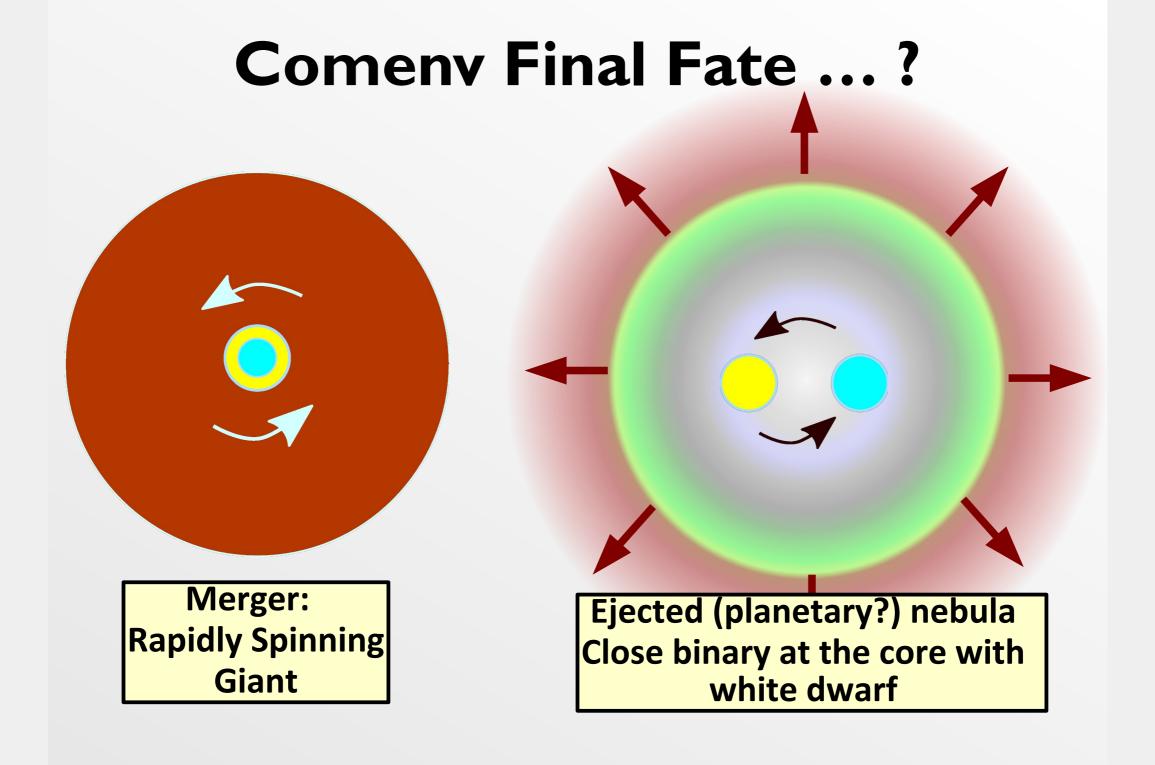
Data from
Giuricin, G.; Mardirossian, F.; Mezzetti
Astronomy and Astrophysics 134, 365

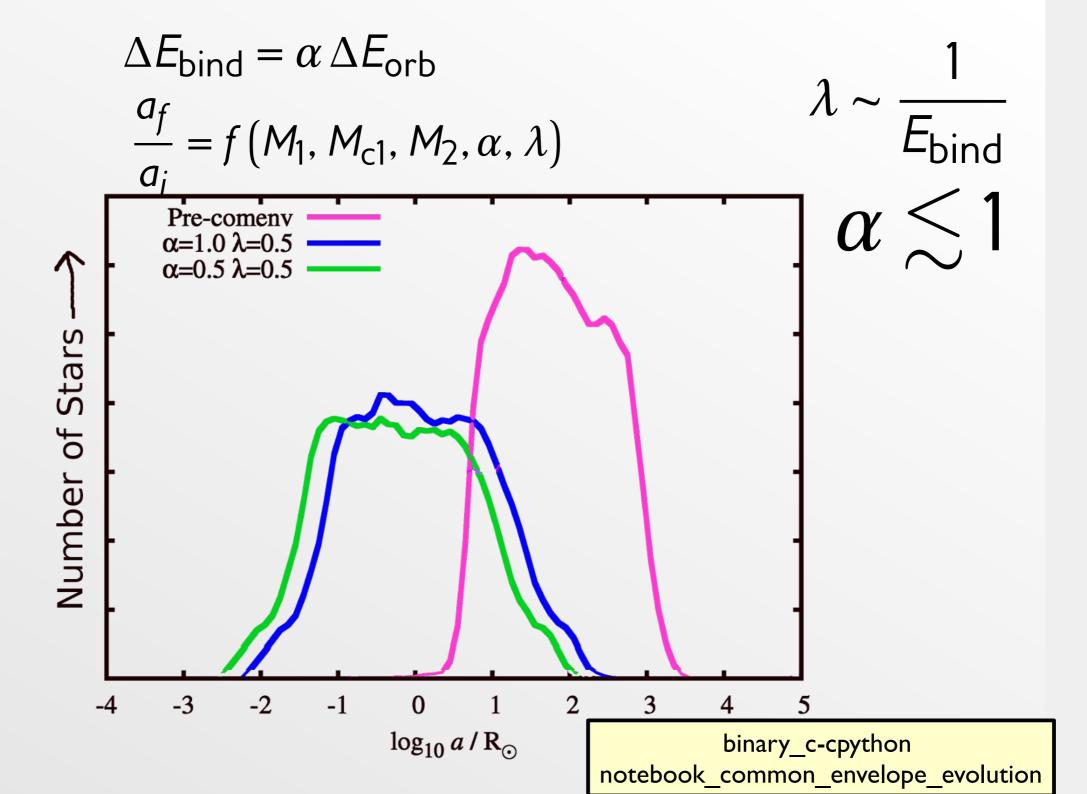




Conservative mass transfer

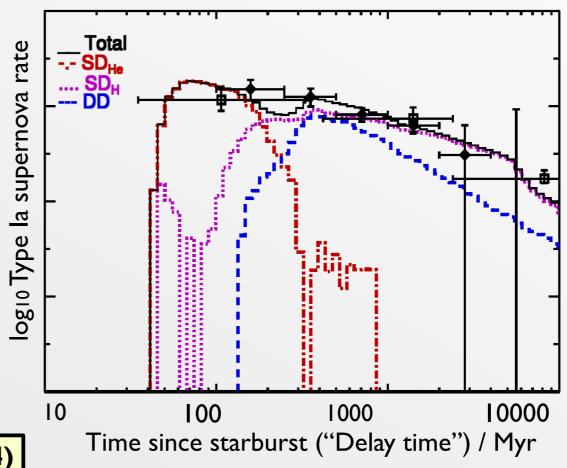






Close systems 1: low/intermedate mass

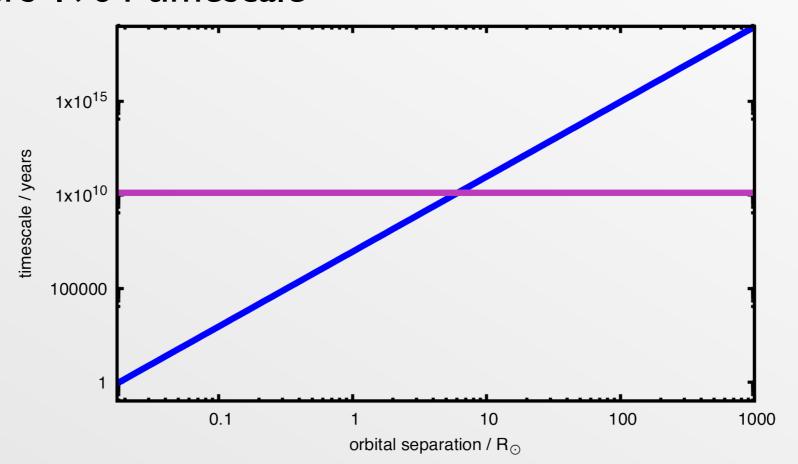
- Donors are usually red-giant stars
 - → hydrogen shell burning, maybe AGB stars → comenv
- → He/CO/ONeWDs, perhaps sdB/O stars
- accretion from
 surviving star →
 novae, type Ia SNe
- COWD? COWD
 binaries → SNe Ia



Claeys et al. (2014)

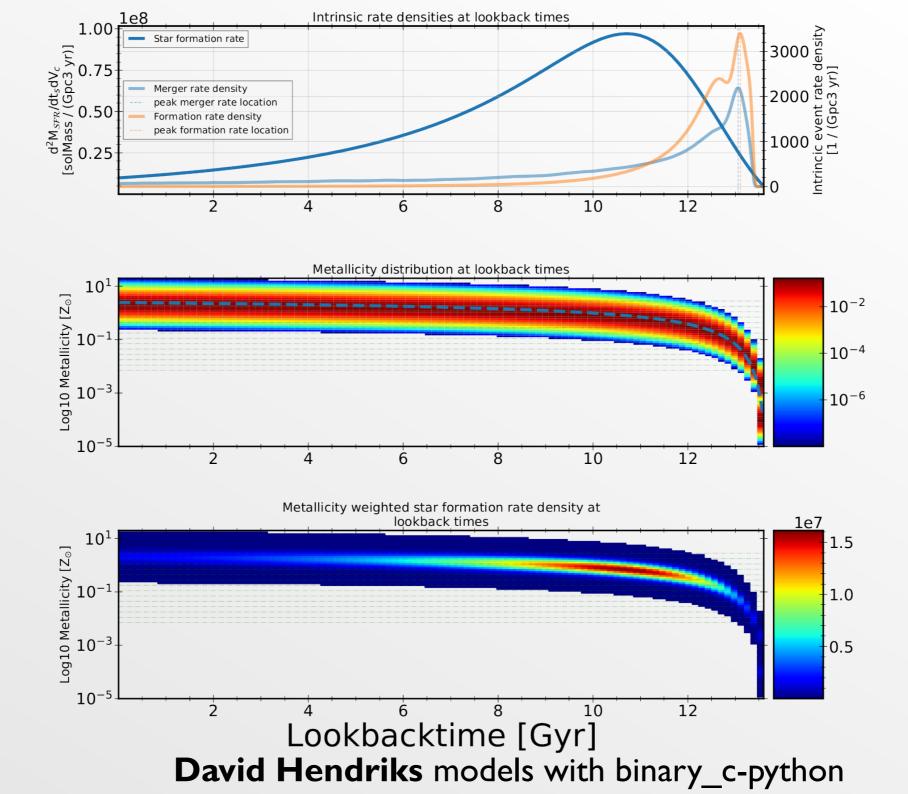
Close systems II: high mass

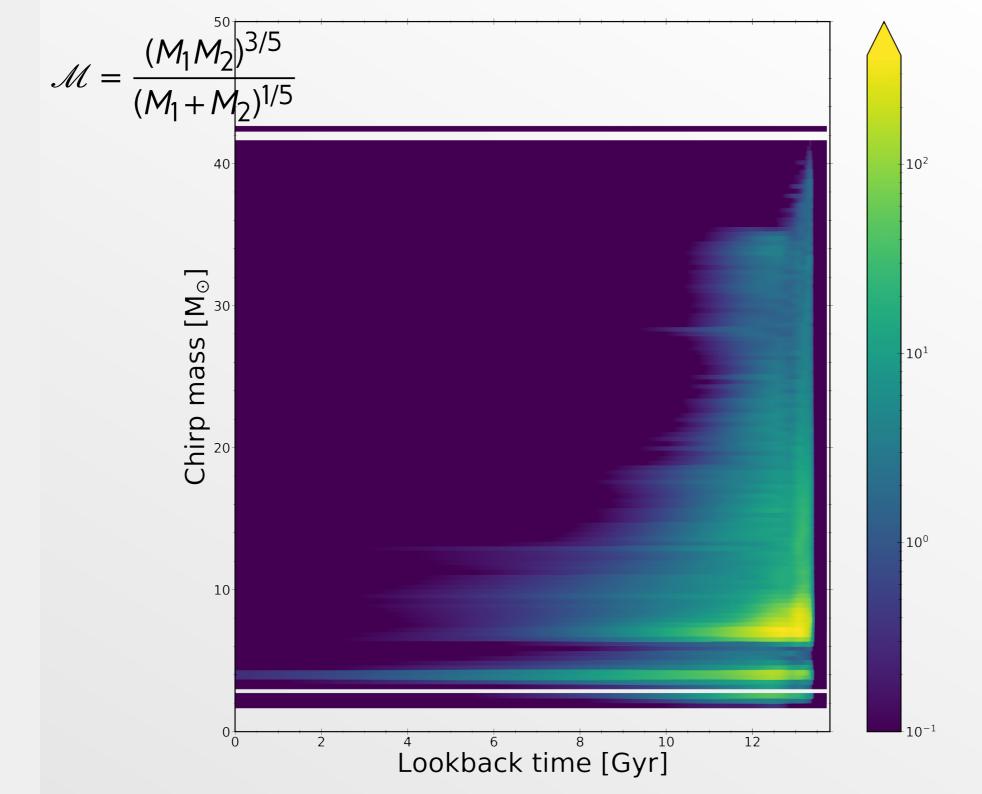
- X-ray binaries → NS+NS, NS+BH, BH+BH systems
 - → close enough for gravitational radiation
- Peters 1964 timescale

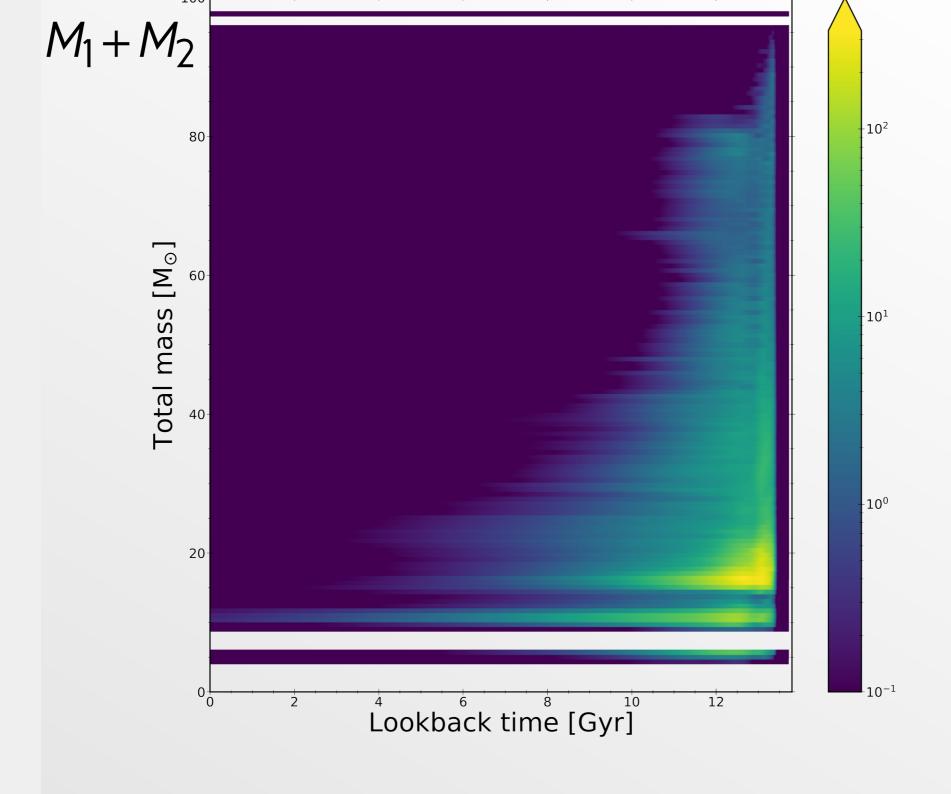


Close systems II: high mass

- X-ray binaries, NS+NS, NS+BH, BH+BH systems
 - → close enough for gravitational radiation
- System needs to survive mass loss and SN kicks
 - → unlikely, so requires "luck" → many systems

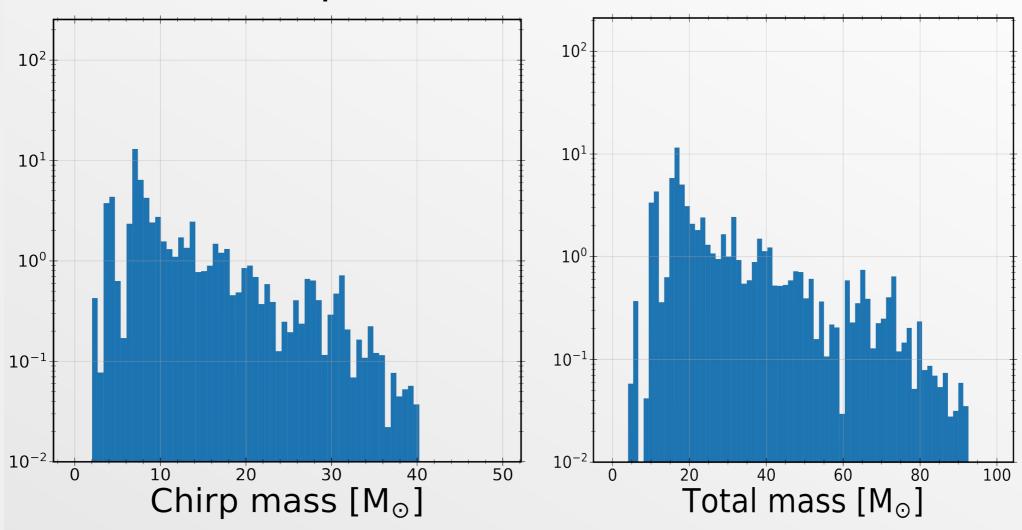






Predicted distributions of BHBH systems now

- → ready to fold in selection effects
- → then compare with observations

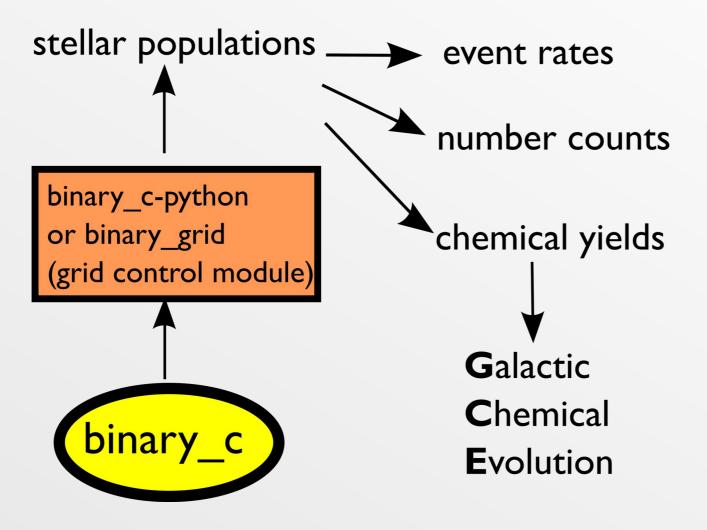


Other important binary physics

- Supernova kicks: e.g. eccentric X-ray binaries
- Exotic stars e.g. helium stars, sdB/O stars
- Gravitational radiation: spiral in of very close binaries:
 NS/BH+NS/BH,WD+WD → bang!
- Rejuvenation: accretion makes stars younger
- Wind accretion in wide binaries (see case studies)
- Circumbinary discs → eccentricity pumping, accretion

All this physics is in my binary popsyn code binary_c http://personal.ph.surrey.ac.uk/~ri0005/binary_c.html

binary_c code



You'd be a mug not to use it!



http://personal.ph.surrey.ac.uk/~ri0005/binary_c.html

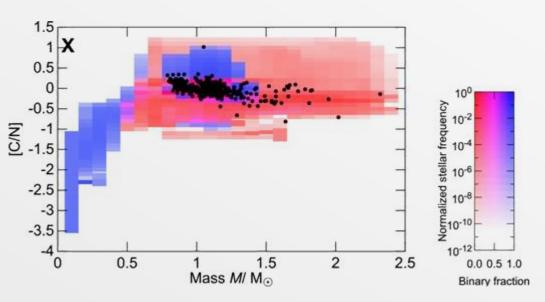
Case studies with binary_c

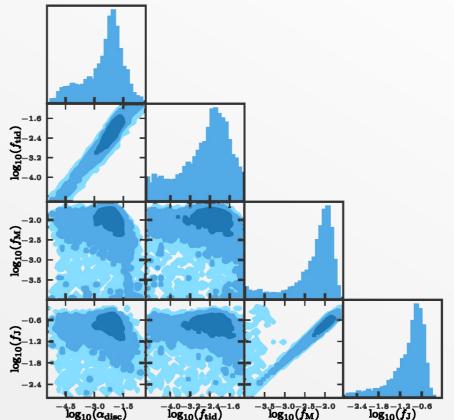
- Massive stars: spin rates, evolutionary outcomes (Sana et al. 2012, de Mink et al. 2013, 2014)
- Mass functions with binary stars (Schneider et al. 2014, 2015)
- Mergers and R-type carbon stars
 (Izzard et al. 2007, Xiang and Jeffery 2013)
- Type la supernovae (Claeys et al 2013, 2014)
- Wind accretion and carbon-rich halo stars (Izzard et al 2009, Abate et al. 2013, 2014, 2015)

Case studies with binary_c

Abundances in merged thick disc stars

(Izzard et al. 2018)





Nova rates in the Milky Way and M31

(Kemp et al. 2021)

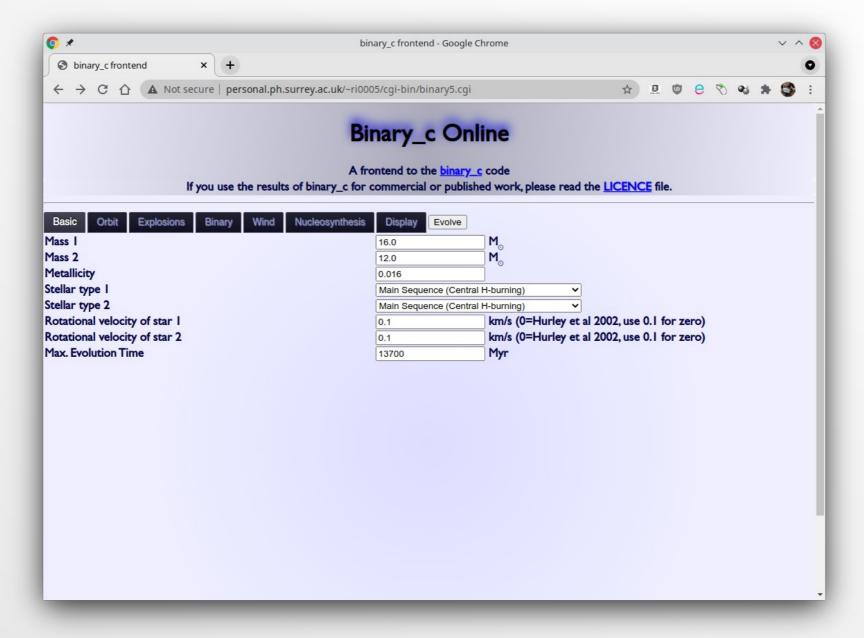
Circumbinary discs around post-AGB stars

(Izzard and Jermyn 2018, 2021)

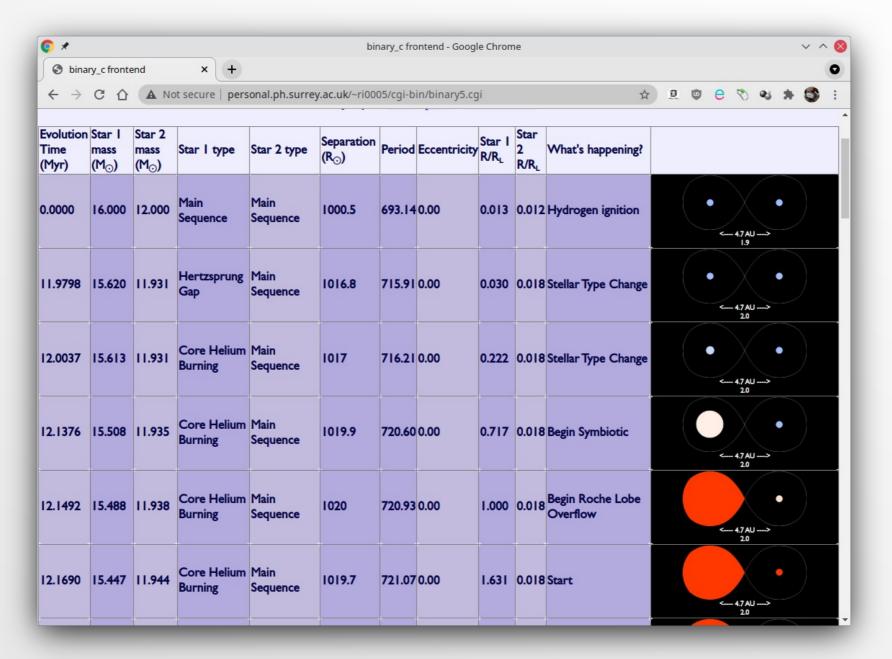
Tides and the MINT library

(Mirouh et al. 2021)

http://personal.ph.surrey.ac.uk/~ri0005/cgi-bin/binary5.cgi



http://personal.ph.surrey.ac.uk/~ri0005/cgi-bin/binary5.cgi



binaryc-python module

- Total control over spacing/distributions
- Multi-D parameter space handled automatically
- loops over single/binary stars, M_1 , M_2 , a, e, whatever!
- Multi-CPU or distributed over computer cluster

```
population.add_grid_variable(
    name="M_1",
    longname="Primary mass",
    valuerange=[1,10],
    resolution=100,
    spacingfunc="const(1,10,100)",
    precode="M_1=math.exp(lnm1)",
    probdist="three_part_powerlaw(M_1, 0.1, 0.5, 1.0, 150, -1.3, -2.3, -2.3)*M_1",
    dphasevol="dlnm1",
    parameter_name="M_1",
)
population.evolve()
```



https://gitlab.eps.surrey.ac.uk/ri0005/binary_c-python Docs at https://ri0005.pages.surrey.ac.uk/binary_cpython/index.html

← David Hendriks main binary_c-python developer 49

For this school

- http://personal.ph.surrey.ac.uk/~ri0005/ Heidelberg2021/school.html
- Installation is either FAST with **Virtualbox**



or SLOW: you build the code yourself



- but you learn how to do it, vital skills!
 - → try both? neither is hard!
- works best on Linux (e.g. Ubuntu) → simple install script
- ... but compiles on OSX and Windows (Linux subsystem)

binary_c-python notebooks

• In the examples/ directory

https://gitlab.eps.surrey.ac.uk/ri0005/binary_c-python/-/tree/master/examples

With documentation

https://ri0005.pages.surrey.ac.uk/binary_c-python/example_notebooks.html

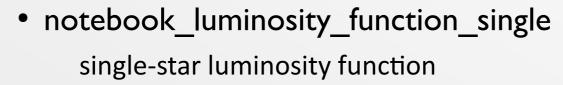
- Tutorials from David
 - Running individual systems with binary_c-python
 - Using custom logging routines with binary_c-python
 - Running populations with binary_c-python
 - Extra features and functionality of binary_c-python
- Use these to really understand how it works
 - → and for vital technical know-how for you!
- notebook_individual_systems is particularly useful to explore how one system evolves and interact

binary_c-python notebooks

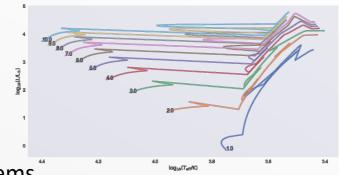
Physics examples from me:

notebook_HRD

make Hertzsprung-Russell diagrams of single/binary systems



- notebook_luminosity_function_binaries binary-star luminosity functions (cf. Schneider+ 2014)
- notebook_common_envelope_evolution
 shows how the orbit changes during common envelope
- notebook_solar_system
 put some rocks in orbit around the Sun... see what happens!
- notebook_BHBH
 BHBH merging systems work in progress, please watch for updates



The future of population synthesis

- Computing speed increase: use detailed models?
- Hybrids may be better: computers have a long way to go to be $> 10^6$ times faster
- Progress in statistics, but how we compare huge parameter spaces and datasets (e.g. Gaia) is still a challenge. Machine learning the way forward? Maybe...
- Need smart people!



