Stellar Tidal Disruption Events with Abundances and Realistic Structures (STARS):
Library of Fallback Rates

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Abstract

We present the STARS library, a grid of tidal disruption event (TDE) simulations interpolated to provide the mass fallback rate (dM/dt) to the black hole for a main-sequence star of any stellar mass, stellar age, and impact parameter. We use a one-dimensional stellar evolution code to construct stars with accurate stellar structures and chemical abundances, then perform tidal disruption simulations in a three-dimensional adaptive-mesh hydrodynamics code with a Helmholtz equation of state, in unprecedented resolution: from 131 to 524 cells across the diameter of the star. The interpolated library of fallback rates is available on GitHub (github.com/jamiealaw-smith/STARS_library) and version 1.0.0 is archived on Zenodo; one can query the library for any stellar mass, stellar age, and impact parameter. We provide new fitting formulae for important disruption quantities (βcrit, ΔM, Mpeak, tpeak, n) in a function of stellar mass, stellar age, and impact parameter. Each of these quantities varies significantly with stellar mass and stellar age, but we are able to reduce all of our simulations to a single relationship that depends only on stellar structure, characterized by a single parameter $\rho_c/\rho_s$ and impact parameter $\beta$. We also find that, in general, more centrally concentrated stars have steeper $dM/dt$ rise slopes and shallower decay slopes. For the same $\Delta M$, the $dM/dt$ shape varies significantly with stellar mass, promising the potential determination of stellar properties from the TDE light curve alone. The $dM/dt$ shape depends strongly on stellar structure and to a certain extent stellar mass, meaning that fitting TDEs using this library offers a better opportunity to determine the nature of the disrupted star and the black hole.

Unified Astronomy Thesaurus concepts: Black hole physics (159); Active galaxies (17); Galaxy nuclei (609); Gravitation (661); Hydrodynamics (1963); Main sequence stars (1000); Tidal disruption (1696)

1. Introduction

Near the region of influence of a galactic massive black hole (BH), a star scattered onto an orbit that brings it close to the BH can be tidally disrupted by the BH’s gravitational field—this is a tidal disruption event (TDE). The stellar material that remains bound to the BH produces a flare that is the signature of this event. TDEs were first explored theoretically with pioneering studies by Lidskii & Ozernoi (1979), Hills (1975), Carter & Luminet (1983), Rees (1988), and Evans & Kochanek (1989).

Several dozen such flares have been observed at the centers of other galaxies (for reviews of observations, see, e.g., Komossa 2015; Auchettl et al. 2017; Hung et al. 2017; van Velzen et al. 2020), with observations now regularly capturing both the rise (e.g., Holoien et al. 2019b) and decay (e.g., Holoien et al. 2018; van Velzen et al. 2019) of the transient in great detail, and even signatures of an accretion disk (e.g., Holoien et al. 2019a; Hung et al. 2020). Fitting theoretical models to observed TDEs allows one, in principle, to extract the properties of the disruption: BH mass, BH spin, stellar mass, stellar age, impact parameter, and radiative efficiency. The first attempt to systematically extract BH masses from TDEs (Mockler et al. 2019) was remarkably successful, obtaining errors of order that of the $M-\sigma$ relationship. However, determination of other parameters is made difficult by degeneracies between stellar properties and BH properties. One needs better theoretical models of TDEs in order to extract much more accurate information from observed events.

A combination of detailed theoretical modeling and high-resolution observations can turn TDEs into unique tools to probe several astrophysical questions: (1) the BH mass function and in particular the possible existence of a cutoff in the BH mass function at low masses ($M_{BH} \lesssim 10^{5}M_{\odot}$); (2) the BH spin distribution; (3) the radiative efficiency of BH accretion and other questions of accretion physics; (4) the stellar populations (stellar masses and ages) in galactic centers, as the stars at the centers of distant galaxies are exposed through their disruption and accretion; (5) the dynamics operating in galactic centers, e.g., which mechanisms (two-body, resonant relaxation, secular effects, etc.) dominate how stars and BHs interact.

At the order-of-magnitude level, tidal disruption occurs when a star crosses the tidal disruption radius

$$r_t = \left( \frac{M_{BH}}{M_*} \right)^{1/3} R_* \propto \tilde{p}^{-1/3},$$

where $p_t = r_t / \tilde{p}$ is the ratio of the tidal radius to the pericenter distance; it is an order-of-magnitude measure of the “strength” of the tidal
interaction. The critical impact parameter $\beta_{\text{crit}}$ is defined as the smallest impact parameter of full disruption (i.e., where the entire star’s mass is disrupted), below which are partial disruptions and above which are “post-critical” encounters.

The mass fallback rate $dM/dt \equiv M_{\text{fb}}$ of debris to pericenter is a central quantity of interest as it appears to track the optical/UV luminosity evolution of observed TDEs closely (e.g., Gezari et al. 2012; Guillochon et al. 2014; Mockler et al. 2019). For nonrelativistic disruptions, one can scale mass fallback rate and time of return to pericenter with BH mass as constant fraction of the star

$$M_{\text{peak}} \propto M_{\text{BH}}^{-1/2} M_{*}^{2} R_{*}^{-3/2}$$

(3)

$$t_{\text{peak}} \propto M_{\text{BH}}^{1/2} M_{*}^{-1} R_{*}^{3/2},$$

(4)

where the $M_{\text{peak}} \propto M_{*}^{2}$ scaling results when we assume that a constant fraction of the star’s mass is lost to the BH in the disruption. Note that these formulae are for the peak quantities of the mass fallback rate, but can be applied to scale the entire $dM/dt$ curve.

Because the tidal radius depends inversely on the average density of the star (Equation (1)), there is a maximum BH mass for disruption outside the innermost-bound spherical orbit for different types of stars. See Figure 1 of Law-Smith et al. (2017a) for the phase space in $M_{\text{yield}}$ and $M_{\text{BH}}$ of stellar objects (from planets to evolved stars) that are expected to produce bright tidal disruption flares. We expect that most observed TDEs will be from MS stars, and that the mass function of TDEs is relatively flat for $M_{*} \lesssim M_{0}$ (Kochanek 2016a) (see the more detailed discussion in Section 4). Thus, it is important to have a library of MS star simulations with which to fit observed events. This work is aimed to be the definitive library of the MS star parameter space. One can extend this library to include relativistic encounters, but these are rarer and account for a small fraction of MS star disruptions (see discussion in Section 4).

Our own Galactic center’s nuclear star cluster is host to young, massive stars, giant-branch stars, and an old population of main-sequence (MS) stars (e.g., Schödel et al. 2007). Stars in the Galactic center also exhibit a range of metallicities, from metal-rich to metal-poor (Feldmeier-Krause et al. 2017). In nuclear clusters outside our galaxy, there is evidence for a diversity of stellar ages and types (Seth et al. 2010) and a wide range of star-formation histories (Georgiev & Böker 2014). Additionally, TDEs appear to be observed preferentially in post-starburst galaxies (Arcavi et al. 2014; French et al. 2016; Law-Smith et al. 2017b; Graur et al. 2018). Thus, we expect some TDEs to be sourced by more massive stars and also by stars of varying ages and compositions. Observations of TDEs suggest that many disruptees are non-zero-age MS (ZAMS) stars (spectra show metal lines that evolve with time; e.g., Leloudas et al. 2019) as is expected (this was first argued for in Kochanek 2016b). However, the theoretical modeling of TDEs thus far has largely ignored the stellar evolution aspect of the problem.

A brief (incomplete) summary of recent theoretical work, where we highlight features relevant to or differing in significant ways from this work, is below. Before discussing the simulation work, we mention the so-called “frozen-in” approximation, studied in detail by Lodato et al. (2009), in which the star arrives intact to pericenter. One can then integrate across the star in equal-orbital-energy slices and determine the mass fallback rate to the BH as a function of time analytically. In this framework, only the structure of the star matters. This approximation can only handle full disruptions (a single impact parameter). Kochanek (2016b) investigated abundance anomalies using this framework and one-dimensional (1D) stellar evolution models from MESA. Gallegos-García et al. (2018), following the above two works, developed an analytic framework to study the composition as a function of time following disruption, using MESA stars. In a recent significant extension to Lodato et al. (2009), Coughlin & Nixon (2019) developed an analytic framework to determine the asymptotic power-law slope of the fallback rate, finding $\propto t^{-5/3}$ for full disruptions and $\propto t^{-9/4}$ for partial disruptions, in general (but not detailed) agreement with simulations.

Despite the success and usefulness of the “frozen-in” approximation in calculating TDE properties, we have learned from hydrodynamical simulations that the star arrives at pericenter significantly distorted and spinning (for a detailed discussion see Steinberg et al. 2019). Nolthenius & Katz (1982) performed the first three-dimensional (3D) tidal disruption simulations, of a $\gamma = 5/3$ polytropic stellar structure. Khokhlov et al. (1993a, 1993b) were the first to study stars with different stellar structures ($\gamma = 5/3, 1.5, 4/3$) numerically. Ramirez-Ruiz & Rosswog (2009) studied the tidal disruption and initial disk formation for $\gamma = 5/3$ and $\gamma = 1.4$ polytropes in 3D hydrodynamics. Guillochon & Ramirez-Ruiz (2013) performed the first systematic parameter-space study of the effect of both stellar structure and impact parameter, exploring $\gamma = 4/3$ and $\gamma = 5/3$ polytropes and a wide range of impact parameters. The effect of impact parameter made it clear that tidal disruptions are a 3D nonlinear hydrodynamical problem that needs to be studied by simulations, at least to a certain extent. Laguna et al. (1993) were the first to explore relativistic tidal disruptions on a Schwarzschild metric. Diener et al. (1997) studied the disruption of a $\gamma = 5/3$ polytrope by a rotating BH on a Kerr metric. Haas et al. (2012) simulated the disruption of a white dwarf by a spinning BH. Cheng & Bogdanović (2014) and Tejeda et al. (2017) compared relativistic simulations to Newtonian simulations in detail, finding good agreement for nonrelativistic encounters. Gafton & Rosswog (2019) performed a grid of general-relativistic simulations for a $\gamma = 5/3$ polytrope, exploring higher impact parameters and spinning BHs, providing new fitting formulae for the relativistic regime, and again finding good agreement for nonrelativistic encounters. See also Stone et al. (2019) for a review of the status of TDEs in general relativity. Rosswog et al. (2008a, 2008b, 2009) studied the tidal disruption of white dwarfs in detail for the first time, while MacLeod et al. (2012) were the first to study the tidal disruption of giant stars, whose highly segregated density profiles did not allow them to be fully disrupted. Law-Smith et al. (2017a) performed a case study simulating realistic stellar structures and compositions, with a hydrogen-envelope helium-core white dwarf. Golightly et al. (2019a) studied the stellar spin dependence of fallback rates for a $\gamma = 5/3$ polytrope. Goicovic et al. (2019) ran moving-mesh simulations of a ZAMS $1 M_{\odot}$ star with a MESA stellar structure. Their results were consistent with the $\gamma = 4/3$ result from Guillochon & Ramirez-Ruiz (2013).

\footnotesize{In the present model fitting, this is true provided a few additional free parameters (mostly related to the photosphere size).}
Golightly et al. (2019b) ran simulations with MESA stellar structures, for three stellar masses and ages, at one impact parameter $\beta = 3$, and argued that the inferred BH mass from fitting TDEs to polytropical simulations can be incorrect at the order-of-magnitude level. Law-Smith et al. (2019) performed simulations with MESA stellar structures and a Helmholtz equation of state (EOS), tracking chemical abundances for 49 elements in the 3D hydrodynamical simulations for the first time. They found significant differences from the polytropic results of Guillochon & Ramirez-Ruiz (2013) and also significant differences from the analytic predictions of the fallback-rate composition of Gallegos-Garcia et al. (2018). Ryu et al. (2020a, 2020b, 2020c, 2020d) recently posted results of a parameter-space study using a fully general-relativistic framework, for a range of stellar masses, impact parameters, and BH masses, and at a single stellar age (see Sections 2 and 4 for comparisons).

A number of theoretical studies have focused on the disk-formation process (e.g., Ayal et al. 2000; Hayasaki et al. 2013; Guillochon et al. 2014; Dai et al. 2015; Guillochon & Ramirez-Ruiz 2015; Piran et al. 2015; Shiokawa et al. 2015; Bonnerot et al. 2016; Hayasaki et al. 2016; Lu & Bonnerot 2020; Bonnerot & Lu 2020). At present, the emission mechanism(s) responsible for the optical/UV emission in TDE flares is an open question—the main candidates are (1) rapid disk circularization and accretion emission or (2) stream self-intersections and disk-formation-process emission—but it is clear that the optical/UV luminosity evolution of TDEs (the “light curve”) tracks the mass fallback rate to the BH ($dM/dt$, the main output of the simulations in this work) very closely (see references above). This means that the fallback rate, which, as we show in this paper, is nearly solely determined by stellar structure, is a useful output.

In this paper, we study the disruption of MS stars in a grid of stellar mass, stellar age, and impact parameter. Previously published systematic studies have used polytropical stellar structures, where $P = K_1\rho^n = K_2\rho^{(n+1)/n}$. This work is a parameter space study using realistic internal stellar structures, chemical abundances, and EOS. We significantly expand upon the stellar structure study of Guillochon & Ramirez-Ruiz (2013) by considering a wide range of density profiles derived from more accurate stellar models (as compared to polytropes): we study 14 distinct stellar structures, corresponding to different stellar masses and ages. The EOS is incorporated via the Helmholtz EOS, consistent with the MESA EOS of the stellar models. We find (see the Appendix) that the EOS contribution to the pressure support is small, and we argue that one can predict many of the properties of tidal disruption from stellar structure and impact parameter alone. In order to reduce our simulations into one relationship for various tidal disruption quantities, we parameterize the stellar structure by the single parameter $\rho_1/\rho$, the ratio of the star’s central density to its average density. The simulations presented in this work have a full Helmholtz EOS tracking 49 elements; however, we study the chemical abundance of the fallback debris in a follow-up paper.

This paper is organized as follows: Section 2 describes our methods, Section 3 describes our results, Section 4 concludes, and the Appendix describes the interpolated STARS library tool, as well as several other issues.

### 2. Methods

We use the 1D stellar evolution code MESA ( Paxton et al. 2011) to run a grid of models from $0.1M_e$ to $10M_e$, from pre-MS to terminal-age MS (TAMS). We define TAMS as a central hydrogen mass fraction of $10^{-3}$. We use the mesa_49.net nuclear network, including 49 elements. See also Gallegos-Garcia et al. (2018) and Law-Smith et al. (2019) for details on the MESA setup. Table 3 in the Appendix lists relevant or non-standard parameters for the MESA simulations. Stars with $M_e \lesssim 0.8M_e$ have an MS lifetime longer than the age of the universe; thus, for these stars, the oldest model we use in our tidal disruption calculations is at 10 Gyr, rather than TAMS.

We then map the 1D stellar density profiles and chemical abundances into the 3D adaptive-mesh refinement (AMR) hydrodynamics code FLASH (Fryxell et al. 2000). We use the Helmholtz EOS and an extended Helmholtz table of terms that spans $10^{-12} \lesssim \rho [g cm^{-3}] \lesssim 10^{15}$ and $10^3 \lesssim T[K] \lesssim 10^{13}$. This Helmholtz EOS is the backbone of the EOS module in MESA, so our mapping is self-consistent. In FLASH, we use the sinks module to integrate trajectories and to track the position of the BH relative to the star, as in Guillochon & McCourt (2017). This is also an update from Guillochon & Ramirez-Ruiz (2013). We begin the simulations at $r = 10r_e$, where tidal effects are negligible.\(^7\) In the multipole gravity solver, we use a maximum angular number of the multipole expansion of $l_m = 20$. Our 3D box is $1000R_e$ on a side and we set the background density to $10^{-11}g cm^{-3}$. This is such that the mass of the box is typically $\ll 0.1%$ of the stellar mass. We use the following hydrodynamics parameters: the interpolation is third order, the slope limiter is the “hybrid” one, and we use the hybrid Riemann solver. We have no magnetic fields ($B = 0$). As in Guillochon & Ramirez-Ruiz (2013), in calculating the binding energy of each cell in the simulation with respect to the BH and stellar debris, we shift the center of mass of the star to be on a parabolic orbit (this is equivalent to centering the $dM/de$ distribution at $e = 0$). Note that in this paper $e$ denotes the specific orbital energy and not eccentricity. Guillochon & Ramirez-Ruiz (2013) show that, because of the magnitude of this shift, it only affects the $dM/dt$ for $t > 100$ yrs. For the vast majority of events, the star is expected to approach on a parabolic or nearly parabolic orbit (Hayasaki et al. 2018). See also Guillochon et al. (2009, 2011), Guillochon & Ramirez-Ruiz (2013), and Law-Smith et al. (2019) for details on the FLASH setup. Table 4 in the Appendix lists several additional FLASH parameters, including the background grid values.

The simulations in this paper are run with $1.5 \times 10^5$ maximum cells. This is a higher maximum cell count than in Law-Smith et al. (2019). We choose the maximum initial refinement based on the central concentration of the stellar density profile. We use 131 cells across the diameter of the star for less centrally concentrated stars ($\rho/\rho_c \lesssim 150$) and 524 cells across the diameter of the star for more centrally concentrated stars ($\rho/\rho_c \gtrsim 150$). In a few test simulations, results are nearly identical if we use 262 or 524 cells across the diameter of the star for the less centrally concentrated stars as well. See Appendix I for our numerical convergence study. For comparison, the simulations in Guillochon & Ramirez-Ruiz (2013) had $\approx 50$ cells across the initial diameter, so our

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\(^6\) Available at [http://cococubed.asu.edu/code_pages/eos.shtml](http://cococubed.asu.edu/code_pages/eos.shtml).

\(^7\) For comparison, Goicoechea et al. (2019) begin at $r = 5r_e$. 

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Law-Smith et al.
simulations have a factor of $2.6 \times 10.5 \times$ higher initial resolution. The moving-mesh simulations of Goicoechea et al. (2019) have a maximum of $2.4 \times 10^5$ cells, thus initially $\approx 60$ cells across the diameter of the star. The simulations of Ryu et al. (2020a, 2020b, 2020c, 2020d) initially have $\approx 50$ cells across the diameter of the star and, in an important difference, the final box size is $17 \times 9 \times 14R_*$. Table 1 lists the parameter space in stellar mass, stellar age, and impact parameter studied in this work. For each star, we run a range of impact parameters $\beta$ (the ratio of the tidal radius to the pericenter distance) from grazing partial disruptions to post-critical disruptions. The lowest $\beta$ for a given star corresponds to $1\%$–$10\%$ mass lost (unbound) from the star in the encounter. We run one very-post-critical (post-full-disruption) $\beta$ for each star such that $r_p = 10r_g$ for this highest $\beta$ encounter, where $r_g \equiv GM/c^2$ in this paper. This is for the purpose of interpolating our grid of $dM/\,dt$ curves in the maximum applicable range. We also plan to extend this library with relativistic simulations (applicable to the small fraction of very relativistic encounters) in future work. For comparison, the most relativistic encounter in Guilliochon & Ramirez-Ruiz (2013) was the $\beta = 4$ for the $\gamma = 4/3$ star, which was $r_p = 11.8r_g$.

Table 1 also lists several other quantities, such as the ratio of the star’s central density to average density $\rho_c/\rho$, a parameterization of the central concentration of the star that we use extensively later on), the dynamical time of the star $t_{\text{dyn}}$, and the corresponding $r_p/r_g$ for each $\beta$.

| $M_*$ | $R_*$ | Stellar Age | $\rho_c/\rho$ | $t_{\text{dyn}}$ | $\beta \equiv r_p/r_g$ | $r_p/r_g$ |
|-------|-------|-------------|---------------|----------------|----------------------|-----------|
| 0.1 $M_\odot$ | 0.1214 $R_\odot$ | 0 Gyr | 5.5 | 213 s | 0.5–1.2 | 25–10 |
| 0.1215 $R_\odot$ | 5.5 | 213 s | 0.5–1.2 |
| 0.3 $M_\odot$ | 0.2814 $R_\odot$ | 0 Gyr | 5.8 | 434 s | 0.6, 0.7, 0.8, 0.9, 1.0, 2.0 | 33, 28, 25, 22, 20, 10 |
| 0.2989 $R_\odot$ | 5.8 | 475 s | 0.6, 0.7, 0.8, 0.9, 1.0, 2.1 |
| 0.5 $M_\odot$ | 0.4452 $R_\odot$ | 0 Gyr | 11 | 669 s | 0.6, 0.8, 1.0, 1.15, 1.4, 2.6 | 44, 33, 26, 23, 19, 10 |
| 0.4564 $R_\odot$ | 12 | 715 s | 0.6, 0.8, 1.0, 1.2, 1.4, 2.8 |
| 0.7 $M_\odot$ | 0.6485 $R_\odot$ | 0 Gyr | 23 | 994 s | 0.75, 1.0, 1.25, 1.5, 3.4 | 46, 34, 28, 23, 10 |
| 0.6793 $R_\odot$ | 10 Gyr | 36 | 1065 s | 0.8, 1.0, 1.15, 1.3, 1.5, 1.7, 3.6 | 45, 36, 31, 28, 24, 21, 10 |
| 1.0 $M_\odot$ | 0.9012 $R_\odot$ | 0 Gyr | 42 | 1362 s | 1.0, 1.25, 1.5, 1.75, 2.0, 4.2 | 42, 34, 28, 24, 21, 10 |
| 1.0455 $R_\odot$ | 4.8 Gyr | 138 | 1702 s | 1.0, 1.5, 2.0, 2.5, 3.0, 3.5, 4.9 |
| 1.2872 $R_\odot$ | 8.4 Gyr | 756 | 2325 s | 1.0, 1.5, 2.0, 3.0, 4.0, 4.5, 5.0, 6.0 |
| 1.5 $M_\odot$ | 1.6275 $R_\odot$ | 0 Gyr | 128 | 2699 s | 1.0, 1.5, 2.0, 2.75, 6.7 |
| 2.0805 $R_\odot$ | 2 Gyr | 1697 | 3901 s | 2.0, 4.0, 6.0, 8.6 |
| 3.0 $M_\odot$ | 1.8896 $R_\odot$ | 0 Gyr | 73 | 2387 s | 0.6–4.0 |
| 3.3192 $R_\odot$ | 0.3 Gyr | 1198 | 5558 s | 1.5, 2.0, 3.0, 4.0, 4.5, 5.0, 7.0, 10.8 |
| 10 $M_\odot$ | 3.6870 $R_\odot$ | 0 Gyr | 38 | 3564 s | 1.0–4.2 |
| 8.4252 $R_\odot$ | 0.02 Gyr | 1292 | 12307 s | 1.5–10.8 |

Note. The simulations in this grid are interpolated in each of these dimensions in order to produce the library. Scaled results are indicated by $\gamma$ for polytropes and $\tau$ for our simulations (see text). The ratio of the central to average density $\rho_c/\rho$, the dynamical time of the star $t_{\text{dyn}}$, and the corresponding $r_p/r_g$ for each $\beta$ is also listed.

We run the simulations to a maximum time of 100$\,t_{\text{dyn}}$ and the star reaches pericenter at approximately 20$\,t_{\text{dyn}}$ into the simulation. For moderate $\beta$s, the debris remains inside the box for 100$\,t_{\text{dyn}}$, but for high $\beta$s the debris begins to leave the box earlier—for the most extreme $\beta$s we study this is at approximately 50$\,t_{\text{dyn}}$. So $dM/\,dt$ distributions are calculated at 30–80$\,t_{\text{dyn}}$ after pericenter. Guilliochon & Ramirez-Ruiz (2013) studied fallback rates at up to 550$\,t_{\text{dyn}}$ after pericenter, showing that the resulting shape is identical to those calculated at earlier times as we do. By not having a large enough box to follow the entire debris for $t > 100t_{\text{dyn}}$, we do not sacrifice any accuracy in the final result, but instead gain precision as a smaller box allows for higher resolution. Note that however, unlike in Ryu et al. (2020a, 2020b, 2020c, 2020d), where debris leaves the computational domain after a few dynamical timescales, the debris in our simulations remains in the box for many dynamical timescales after pericenter. The effect this difference has on fallback-rate determinations is unclear.

As an example of our analysis method, Figure 1 shows the specific binding energy distribution and resulting mass return...
rate to the BH for a $0.3 M_\odot$ ZAMS star in a $\beta = 0.9$ encounter with a $10^6 M_\odot$ BH. $dM/de$ is plotted in units of $M_*/\Delta e$, where $\Delta e = GM_*^2/3 M_{BH}^3/R_*$, an order-of-magnitude estimate of the range in fluid binding energies (Stone et al. 2013). The x-axis is normalized similarly. The hydrodynamical grid data from the simulation is binned and then fit with a B-spline. The $dM/dt$ is extended by finding the average slope of the last 10%–20% of the $dM/dt$. The extended section is not visible on this plot as it is for $t \gtrsim 5\times 10^5$ s. The result for a $\gamma = 5/3$ polytrope, for this $\beta$, scaled to the same mass and radius as this star, is also shown. It is expected to match quite closely as the stellar structure of a $0.3 M_\odot$ ZAMS star is well approximated by a $\gamma = 5/3$ polytrope (see Figure 17).

The small differences between the polytropic and MESA initial condition simulations are likely due to differences in resolution, numerical method, and smoothing algorithm between the two works. The B-splines applied to smooth the $dM/de$ distributions in this work have been examined and calibrated in detail for the $dM/de$ result from each simulation, and so the resulting $dM/dt$ curves in this work have higher fidelity to the raw hydrodynamical grid data.

3. Results

3.1. Stellar Structure

Figure 2 shows the ratio of central density to average density $\rho_c/\bar{\rho}$ for MS stars, interpolated based on a finely spaced MESA grid (more finely spaced than for our TDE simulations; Table 1). This shows the range of stellar structures on the main sequence. We perform TDE simulations in FLASH for structures at the extremes and interpolate the resulting $dM/dt$’s in stellar mass and stellar age in between these extremes. For some regions one can use polytropic stellar structures rather than MESA initial conditions (e.g., for $M_* \lesssim 0.3 M_\odot$).

The top panels show $\rho_c/\bar{\rho}$ in the space of stellar age versus stellar mass. Here one sees that lower-mass stars ($\lesssim 0.8 M_\odot$), whose MS lifetimes are $\gtrsim$ the age of the universe, evolve slowly and have roughly constant stellar structures over 10 Gyr. More massive stars ($\gtrsim 0.8 M_\odot$) have shorter lifetimes and evolve through different stellar structures more rapidly. We expect a wide variety of stellar masses and stellar ages to source observed TDEs; see Section 4 for discussion.

The bottom panels show normalized density profiles colored by $\rho_c/\bar{\rho}$, both individually for the stars we simulate in FLASH and interpolated on a more finely spaced grid of MESA stars. One can see that $\rho_c/\bar{\rho}$ maps nearly 1–1 onto the stellar density profile, at least for the main sequence. This makes it a very good single-value parameter to describe the stellar structure, and we use it to reduce our simulations into a single relationship and to provide fitting formulae for each TDE quantity.

For comparison, for giant stars, $\rho_c/\bar{\rho} \gtrsim 10^6$. In MacLeod et al. (2012), due to numerical limitations, the authors use $\rho_{\text{core}}/\bar{\rho} \approx 2 \times 10^3$ for the hydrodynamical simulations. They are unable to fully disrupt the giant stars, due to the highly segregated density profile (much more centrally concentrated than that of MS stars despite the artificial value of $\rho_{\text{core}}/\bar{\rho}$) and extended envelope structure.

Figure 17 in the Appendix shows density profiles for each star we simulate as compared to $\gamma = 5/3$ and $\gamma = 4/3$ stellar structures. Also see Figure 1 in Law-Smith et al. (2019) for the stellar structure evolution of a $1 M_\odot$ and $3 M_\odot$ star over their MS lifetimes (also compared to polytropic stellar profiles). Only a few stars correspond closely to polytropic stellar structures; for all others, MESA initial conditions are significantly more accurate.

3.2. Qualitative Tidal Disruption Results

As an example of one of our FLASH simulations, Figure 3 shows a volume rendering of the disruption of a $1 M_\odot$ ZAMS star at $t = t_p \approx 3 t_{\text{dyn}}$ in a $\beta = 1$ encounter with a $10^6 M_\odot$ BH (where $t_p$ is the pericenter time and $t_{\text{dyn}}$ is the dynamical time of the star). Color corresponds to density. One can see the stratified structure following the disruption of a star with an accurate internal stellar structure. This encounter is a partial disruption in which a core survives. Note that this snapshot is zoomed in on the star in order to highlight the density structure of the debris a few dynamical times after pericenter, but that
Figure 2. Stellar structure, parameterized by the ratio of central density to average density. Top: as a function of stellar mass and stellar age for MS stars, based on finely spaced MESA grid. x-axis is log scale. y-axis is linear scale on left panel and log scale on right panel and its maximum is 13.8 Gyr. Dotted line corresponds to $\gamma \approx 5/3$ polytrope ($\rho_c/\bar{\rho} = 5.8$, from a ZAMS 0.3 $M_\odot$ star) and dashed line corresponds to $\gamma \approx 4/3$ polytrope ($\rho_c/\bar{\rho} = 73$, from a ZAMS 3 $M_\odot$ star). Black circles are stars for which we have done tidal disruption simulations in FLASH (note we run multiple $\beta$’s for each star). Bottom: density profiles for MS stars, normalized to central density and stellar radius. Left panel shows stars we have disrupted in FLASH (numbers in legend are $\rho_c/\bar{\rho}$) and the right panel is based on a finely spaced MESA grid. Dotted and dashed lines are $\gamma = 5/3$ and $\gamma = 4/3$ polytropes, respectively.

Figure 3. Volume rendering of a 1$M_\odot$ ZAMS star at $t - t_p \approx 3t_{dp}$ in a $\beta = 1$ encounter with a 10$^6M_\odot$ BH. Color corresponds to density, and the color bar and transfer function are chosen to qualitatively highlight the stratified density structure of the debris. Videos of the simulations are available at https://www.youtube.com/channel/UCShahcFYgrjSdOZ7TtvOEqSOA.
our computational domain is 1000\(R_\odot\) on a side (roughly 100 times the size of this volume rendering), and the debris eventually expands to fill this.

As an example of the range of vulnerability to tidal disruption of our grid, Figure 4 shows two-dimensional (2D) density slices of the disruption of several different stars at different impact parameters, all at \(t = 2t_{\text{dyn}}\) after pericenter. The axes are arranged such that \(\beta\) increases from left to right, and central concentration \((\rho_c/\bar{\rho})\) increases from top to bottom. The white contours correspond to absolute values of density \(1, 10^{-1},\) and \(10^{-2}\) \(\text{g cm}^{-3}\). Figure 4. 2D density slices of a few different simulations. The axes of this plot grid are central concentration \((\rho_c/\bar{\rho})\) vs. impact parameter \((\beta/\beta_c)\). Each row is a different star and each panel is a different impact parameter. \(\beta\) increases from left to right and \(\rho_c/\bar{\rho}\) increases from top to bottom. Color is the logarithm of density and is normalized to the initial central density of the star, extending to \(10^{-8}\) of this value. Each panel has width \(10R_\odot\). All panels are at \(t = \approx 2t_{\text{dyn}}\). White contours correspond to densities of \(1, 10^{-1},\) and \(10^{-2}\) \(\text{g cm}^{-3}\).

that are more differentiated. This leads to the outer layers being torqued more than the inner layers, and the core remaining sequestered and undisturbed at higher \(\beta\)'s for more centrally concentrated stars. Increasing central concentration allows the star to survive higher-\(\beta\) encounters. For example, a ZAMS 0.3\(M_\odot\) star has a larger fraction of its mass at larger radii, and is thus fully disrupted by a \(\beta = 1\) encounter, whereas a TAMS 1\(M_\odot\) star has only a small fraction of its mass at larger radii from its sequestered core, and is thus relatively undisturbed by a \(\beta = 1\) encounter.

### 3.3. Mass Lost

Figure 5 shows fractional mass lost from the star \(\Delta M/M_\star\) as a function of impact parameter \(\beta\). The x-axis in the top-left
The bottom panels have a scaled $x$-axis,

$$x = \exp[(\beta / \beta_{\text{crit}})^\alpha - 1], \quad \alpha = (\rho_c / \bar{\rho})^{-1/3},$$

constructed by trial-and-error in order to reduce all of the simulations into a single relationship. The physical meaning of the functional form of $x$ is unclear at present, but the sharp $1/3$ factor suggests at least a partial physically motivated origin as this number appears in basic TDE theory. This formula accounts for the dependence of $\Delta M$ versus $\beta$ on stellar structure. The fact that we are able to express all of the simulations in a single relationship implies that $\Delta M$ depends only on $\rho_c / \bar{\rho}$ and $\beta$. Analytic and B-spline fits for this relation, allowing one to obtain the mass lost for any stellar mass, stellar age, and impact parameter, are provided in the Appendix. The simulations in this work all reduce to a single relation for other disruption quantities as well (see below in this section).

### 3.4. Critical Impact Parameter

We did not sample finely enough in $\beta$ to find the exact critical impact parameter for full disruption for each star, but we estimate this within the bounds associated with our spacing in $\beta$ and the mass lost as a function of $\beta$ for the star. Figure 6 shows the critical $\beta$ as a function of $\rho_c / \bar{\rho}$. For nearly all stars, the critical $\beta$ is well fitted by a simple relation,

$$\beta_{\text{crit}} \approx 0.5 \left(\frac{\rho_c}{\bar{\rho}}\right)^{1/3}, \quad \rho_c / \bar{\rho} \lesssim 500.$$

We note that this formula (for stars with lower central concentrations, $\rho_c / \bar{\rho} \lesssim 500$) is very similar to that found by Ryu et al. (2020b).
At the highest central concentrations ($\rho_c/\bar{\rho} \gtrsim 500$), the critical $\beta$ is higher than predicted with this relation, and is instead better fit by a steeper relation,

$$\beta_{\text{crit}} \approx 0.39 \left(\frac{\rho_c}{\bar{\rho}}\right)^{1.23}, \quad \rho_c/\bar{\rho} \gtrsim 500. \tag{8}$$

Equations (7) and (8) allow one to predict the approximate critical impact parameter for full disruption for any MS star. We note that already for $\rho_c/\bar{\rho} \gtrsim 100$, Equation (7) slightly underpredicts $\beta_{\text{crit}}$ and that there is likely a smooth transition between the two power-law relationships, but we fit only two equations for simplicity. All one needs is $\rho_c/\bar{\rho}$ for that star, obtained from, e.g., a MESA model directly or precomputed grids such as the MIST models (Choi et al. 2016). So the ability to fully disrupt a star is a simple function of this ratio of densities.

Note that for two points at the highest central concentrations, the quoted $\beta_{\text{crit}}$ is obtained through extrapolation, as the maximum $\beta$ simulated for that star did not fully disrupt it. The lower limits obtained from our simulations are also shown. We note also that the resolution (in terms of maximum number of cells in the simulation) required to precisely determine $\beta_{\text{crit}}$ becomes significantly higher for these most centrally concentrated stars. At the end of the simulation when the debris has expanded, and thus the linear resolution has decreased from the initial maximum level of refinement (in order to resolve the same number of cells in the simulation box), the highest density “core” of the star—for the stars with $\rho_c/\bar{\rho} \gtrsim 10^3$—becomes smaller than a grid cell and so is not resolved. However, the aim of this study is not to precisely determine the critical impact parameter for these most centrally concentrated stars, but rather to determine the mass fallback rate to the BH following their disruption.

Table 2 lists the critical impact parameter $\beta_{\text{crit}}$ for full disruption for all stars studied. Errors represent the grid spacing and so are overestimates. We also list the corresponding physical pericenter distance in gravitational radii $r_p/r_g$. For stars that do not evolve significantly in structure over 10 Gyr, their $\beta_{\text{crit}}$ remain the same or very similar. For stars that evolve significantly over <10 Gyr, the $\beta_{\text{crit}}$ increases dramatically over a star’s lifetime. For example, for a 1$M_\odot$ star, the $\beta_{\text{crit}}$ increases from $\approx 1.8$ to $\approx 7.0$, a factor of $3.9 \times$, over its MS lifetime, a result of its $\rho_c/\bar{\rho}$ increasing from 42 to 756. This corresponds to the critical $r_p/r_g$ for full disruption decreasing by a factor of $2.7 \times$, from 23.4 to 8.6. Thus, a TAMS 1$M_\odot$ star must approach $2.7 \times$ closer to the BH in order to be fully disrupted compared to its ZAMS self. Similar trends are seen for the 1.5$M_\odot$ and 3$M_\odot$ stars, where the critical $r_p/r_g$ decreases by a factor $>2 \times$ from ZAMS to TAMS. Thus, for higher-mass stars ($M_* \gtrsim 0.8M_\odot$), at the same pericenter distance, the ZAMS star is roughly $2 \times$ as vulnerable to tidal disruption and associated mass loss as the TAMS star. For lower-mass stars ($M_* \lesssim 0.8M_\odot$), the critical $r_p/r_g$ is roughly constant over 10 Gyr. See the Appendix for the dependence of $\beta_{\text{crit}}$ and the associated $r_p/r_g$ with stellar mass.

As $\rho_c/\bar{\rho}$ increases, the star’s expansion in response to mass loss decreases. As a result, the star becomes less vulnerable to tidal disruption and thus the critical impact parameter for full disruption monotonically increases with increasing $\rho_c/\bar{\rho}$. For example, note the transition in critical impact parameter from $\beta_{\text{crit}} = 0.9 < 1$ for the 0.3$M_\odot$ star to $\beta_{\text{crit}} = 1.1 > 1$ for the 0.5$M_\odot$ star. The 0.3$M_\odot$ star corresponds to a $\gamma = 5/3$ polytrope (see Figure 17) and the 0.5$M_\odot$ star has an intermediate structure in between a $\gamma = 5/3$ and $\gamma = 4/3$ polytrope. In the most extreme cases, for example for giant stars (MacLeod et al. 2012) with a sequestered core-envelope structure, the remaining envelope contracts and thus its disruption requires increasingly higher $\beta$ for the same amount of mass loss (in fact, these stars are unable to be fully disrupted by massive BHs).

3.5. Mass Fallback Rates

3.5.1. All Encounters

Figure 7 shows mass fallback rates $dM/dt$ for all of our simulations, grouped by star. Each panel is a particular stellar mass and age, and shows all impact parameters $\beta$ for that object. One can perhaps pick out trends in the shape of the $dM/dt$ curves that we will explore in more detail below. Higher $\beta$ generally corresponds to shifting the $dM/dt$ curve upwards and
to the left, giving a shorter peak timescale and higher peak fallback rate. After the critical $\beta$ for full disruption, the $\dot{M}/dt$ curve flattens, particularly near the peak. The rise and decay slopes also vary with stellar mass, stellar age, and impact parameter.

The top panels of Figure 8 show $\dot{M}/dt$'s for a single stellar mass, $1 M_\odot$, for all $\beta$'s and stellar ages, in order to demonstrate the effect of stellar age. Note that the $\beta$'s are not the same for each stellar age (see Table 1 or Figure 7). The left panel shows the raw curves and the right panel is normalized to the peak fallback rate ($\dot{M}_{\text{peak}}$) and the associated peak fallback time ($t_{\text{peak}}$) in order to focus on the rise and decay slopes. In the left panel, the curves largely overlap, demonstrating the degeneracy between stellar structure and $\beta$. This is similar to the shifting of $\dot{M}/dt$'s from

Figure 7. Mass fallback rate $\dot{M}/dt$ to the BH as a function of time for all of our simulations. Each panel is a different star (stellar mass and stellar age, labeled in top left). Colors correspond to different impact parameters $\beta$. 
\( \gamma = 5/3 \) and \( \gamma = 4/3 \) polytropes to lie on top of each other demonstrated in Guillochon & Ramirez-Ruiz (2013). This is why there is a degeneracy in the MOSFiT TDE fitting between stellar mass and impact parameter (and so radiative efficiency) (Mockler et al. 2019; Mockler & Ramirez-Ruiz 2020). In the right panel, one sees that the rise and decay slopes also largely overlap for a single stellar mass at multiple ages.

Thus, it is more challenging to identify the age of the star by the light curve alone, due to the degeneracy introduced by \( \beta \). However, stars of different age have significantly different compositions, and thus the stellar debris will have different chemical abundances. This is a promising avenue to distinguish stellar age, and indeed also stellar mass, if one can tie the composition of the fallback debris to observed spectral features. See Law-Smith et al. (2019) for a first study of the chemical abundance of the debris.

The middle and bottom panels of Figure 8 show \( \frac{dM}{dt} \)’s for all of our simulations. The middle panels are colored by stellar mass, with multiple stellar ages and \( \beta \)’s for each mass. From the middle left panel, evidently, more massive stars produce \( \frac{dM}{dt} \) curves with higher normalizations than less massive stars, but the peak timescales are similar if BH mass is constant. This in principle would offer a way to determine stellar mass in an observed TDE, but there is a degeneracy between stellar mass and efficiency in the current fitting of TDE observations (e.g., Mockler et al. 2019; Mockler & Ramirez-Ruiz 2020). One can have more efficient conversion of matter to radiation with a smaller total stellar debris mass, or a less efficient conversion of matter to radiation with a larger total stellar debris mass.

From the middle-right panel, one sees that, generally, the rise slope is shallower with increasing stellar mass and (though this is more difficult to see) the decay slope is steeper with increasing stellar mass. The rise slope is a stronger trend, whereas the decay slope appears to mainly be a function of the fact that for more massive stars, a higher fraction of encounters studied in this work are partial disruptions. Indeed, with lower \( \beta \) encounters being more likely, it is more likely to have a steeper decay slope for a more massive star. We also tabulate the asymptotic power-law indices \( n_{\infty} \) later on in this section.

The bottom panels are colored by \( \rho_\text{f} / \rho \). In the bottom-right panel, one can see that more centrally concentrated stars have shallower rise slopes and (though it is difficult to see because of the density of lines) steeper decay slopes. The fact that more centrally concentrated stars have slower rises can be understood by thinking about (1) the density profile of the star and (2) the hydrodynamics of disruption. First, stars with higher values of \( \rho_\text{f} / \rho \) have relatively more extended, lower density envelopes. Just considering the undisturbed spherically symmetric profile (see Figure 17), the material that will be most bound to the BH has a slower transition in density, leading to a shallower rise slope. Stars with lower values of \( \rho_\text{f} / \rho \), on the other hand, have more uniform density profiles, and their outer layers exhibit a sharper transition in density, leading to a steeper rise slope. Second, there is a hydrodynamical effect that, for a more centrally concentrated star, the outer layers are more vulnerable to tidal disruption and this material is thus further stretched out post-distruption (compare the density contours in Figure 4). This results in a greater difference in rise slopes than might be predicted from the undisturbed density profiles alone. This can be seen more clearly in Figure 9, which shows only full disruptions and has fewer profiles plotted. The fact that more centrally concentrated stars have faster decays relates to the survivability of the core. While full disruptions decay closer to \( \propto t^{-5/3} \), the presence of a surviving core changes the binding energies of the material bound to the BH (Guillochon & Ramirez-Ruiz 2013). This results in a steeper decay slope when there is a surviving core (Coughlin & Nixon 2019, discuss this effect in more detail). See also Ryu et al. (2020c) for further discussion. The transition between partial disruptions and full disruptions occurs at higher \( \beta \)’s with increasing \( \rho_\text{f} / \rho \); thus, more centrally concentrated stars are more likely to exhibit steeper decays.

The structure of the star is imprinted on the shape and slope of the fallback curve. It is thus easier to determine the stellar structure of the disrupted star in an observed TDE than more degenerate properties such as stellar mass and stellar age. This also hints that, in principle, there may be a mapping from a single structural parameter such as \( \rho_\text{f} / \rho \), in combination with \( \beta \), to the shape of the \( \frac{dM}{dt} \). One can imagine a mapping in which \( M_* \) gives the normalization of the \( \frac{dM}{dt} \) while \( \rho_\text{f} / \rho \) gives the “tilt” corresponding to the rise and decay slopes.

### 3.5.2. Full Disruptions

Figure 9 shows only critical (full) disruptions for all stellar masses and ages. These encounters thus all have the same fractional mass lost \( \Delta M / M_* \approx 1 \). Here the trends are more evident than in Figure 8, though the decay slopes are relatively similar, as all full disruptions give \( n_\infty \approx -5/3 \) (see below in the section). The top panels show a single stellar mass \( M_* \approx 1 M_\odot \), in order to highlight the stellar age dependence. Here the actual amount of mass unbound from the star \( \Delta M \) and thus the mass eventually fed to the BH, \( \approx \Delta M / 2 \), is identical between the stars. The rise slopes are shallower with increasing stellar age.

The middle panels are colored by stellar mass. In the left panel one primarily sees the normalization to \( \frac{dM}{dt} \) given by \( M_* \). With increasing \( M_* \), the \( \frac{dM}{dt} \) curve is shifted upwards and slightly to the left. In the right panel, one sees that the rise slopes are shallower with increasing stellar mass. There is a weak trend toward steeper decay slopes with increasing stellar mass.

The bottom panels are colored by \( \rho_\text{f} / \rho \). Here one sees perhaps most clearly the trends discussed above, as we are coloring by the key physical parameter. It appears that, for the same \( \Delta M / M_* \), \( \rho_\text{f} / \rho \) nearly uniquely determines the rise slope of \( \frac{dM}{dt} \) (this is also true for the decay slope, but is more clear when we study \( n_\infty \) versus \( \beta \) below). Increasing central concentration leads to shallower rise slopes, and, for full disruptions, similar decay slopes. See the discussion above for the physical intuition behind this effect.

### 3.5.3. At Fixed Mass Lost

Figure 10 shows a comparison of \( \frac{dM}{dt} \) curves at fixed mass lost \( \Delta M \). This allows one to compare fallback rates for the same amount of total material supplied to the BH. We compare at \( \Delta M = (0.1, 0.3, 0.5, 0.7, 1.0) M_\odot \); half of this unbound material returns to the BH. Of course, some values of \( \Delta M \) are inaccessible to some stellar masses. To do this, we construct a very finely interpolated grid (see the Appendix) of impact parameters for each star, and select the \( \beta \) for which the \( \frac{dM}{dt} \) integrates to the given \( \Delta M \).

The top five panels show stars at 0 Gyr (ZAMS) and the bottom five panels show stars at 10 Gyr (or TAMS for \( M_* \approx M_\odot \)). If the efficiency of conversion of matter to light is relatively constant, then the total energy radiated in a TDE is...
determined by ΔM. Thus, this figure shows dM/dt’s from a stellar population at a single age, and (assuming constant efficiency) for a given radiated energy.

At fixed ΔM, the shape of the dM/dt’s vary significantly with stellar mass. The peak fallback rate, the peak timescale, and the rise and decay slopes all vary significantly. Most strikingly, for...
both stellar age populations, the decay slopes are steeper with increasing stellar mass. This is because for the same $\Delta M$, the encounter is more of a partial disruption for higher stellar masses. Higher stellar masses also correspond to higher $M_{\text{peak}}$ values (by up to a factor of $\approx 5$) at fixed $\Delta M$. The differences in the shape of the $dM/dt$'s are not particularly more or less significant for different values of $\Delta M$. These differences suggest that fitting TDEs with this library of fallback rates will allow for better determination of stellar properties, and perhaps that the nature of the disrupted star can be determined from the light curve alone.

Figure 9. Same as Figure 8, but only critical (full) disruptions. Top: only $M_* = 1M_\odot$, colored by stellar age. Middle: colored by stellar mass. Bottom: colored by $\rho/\bar{\rho}$. 

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3.6. Derived Quantities

3.6.1. Peak Timescale and Fallback Rate

Next we discuss quantities derived from the $M_\dot{M}$ results. The top panels of Figure 11 show the peak mass fallback rate $\dot{M}_{\text{peak}}$ and the associated peak time $t_{\text{peak}}$ as a function of impact parameter $\beta$. The $t_{\text{peak}}$ values are plotted with a linear y-axis and so may appear to vary significantly, but nearly all $t_{\text{peak}}$ values are $\approx 20$–40 days. This is because the mass–radius relationship of MS stars nearly cancels out the $M_*$ and $R_*$ dependence ($M_\ast^{-1}R_*^{3/2}$) in Equation (4). The general trend is that $t_{\text{peak}}$ decreases sharply with $\beta$ until full disruption and then increases more slowly post critical-$\beta$.

The $\dot{M}_{\text{peak}}$ values show a marked $M_*$ dependence, and span several orders of magnitude. We can understand this because of

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{figure10.png}
\caption{Comparison of $\dot{M}/dt$ curves at a fixed mass lost $\Delta M$. The top five panels are stellar ages of 0 Gyr and the bottom five panels are stellar ages of 10 Gyr (or TAMS for $M_* > M_\odot$). These are interpolated $\dot{M}/dt$ curves at the particular $\beta$ for each star that corresponds to the $\Delta M$ in the top left of each panel.}
\end{figure}
the $M_*$ and $R_*$ dependence ($M_1^2 R_1^{3/2}$) in Equation (4) has a higher power on $M_*$ than in the $t_{\text{peak}}$ scaling. The general trend is that $\dot{M}_{\text{peak}}$ increases sharply with $\beta$ until full disruption and then decreases more slowly post critical-$\beta$. Trends for both $t_{\text{peak}}$ and $\dot{M}_{\text{peak}}$ are qualitatively similar to the Guillochon & Ramirez-Ruiz (2013) simulations, but our study spans a wider range due to the many different stellar structures. Focusing on the 1$M_\odot$ star at three ages for comparison, at the same $\beta$, the more centrally concentrated (older) star peaks later and has a lower peak fallback rate.

In the bottom panels of Figure 11, the $x$-axis is scaled with the structural parameter $\bar{r} = -\frac{c_{13}}{r}$ in the same way as in Figure 5. Here $\dot{M}_{\text{peak}}$ is also normalized by the stellar mass. $\dot{M}_{\text{peak}}$ is normalized by the stellar mass. $M_{\text{peak}} \times t_{\text{peak}} / M_* \approx \Delta M_{\text{bound}} / M_* \approx \Delta M / 2M_*$, so this relation is similar to the bottom-left panel of Figure 5. This is a good internal consistency check. Note that the integrated $\dot{M}/dt$ is only approximately equal to the product of its peak values (e.g., for full disruptions $\Delta M / 2M_* = 0.5$, while $M_{\text{peak}} \times t_{\text{peak}} / M_*$ asymptotes to $\approx 0.25$).

3.6.2. Power-law Indices

The top panels of Figure 13 show the instantaneous power-law index $n(t)$ of the mass fallback rate as a function of time. The top-left panel shows the parameter space occupied by tidal disruptions of MS stars for a 10$^7M_\odot$ BH in absolute units. The $x$-axis in the top-right panel is normalized by the peak time, and we also compare to the results of Guillochon & Ramirez-Ruiz (2013) for $\gamma = 5/3$ and $\gamma = 4/3$ polytropes. The regions are comparable, but this work covers more parameter space, as expected by the larger range of stellar structures studied. The different range of $\beta$'s sampled may also account for some differences.

The top panels show the asymptotic decay power-law index $n_\infty$ for individual stars as a function of $\beta$, and as a function of the scaled $x$-axis incorporating stellar structure familiar from earlier figures. Partial disruptions generally have...
stars with exactly $\gamma = 5/3$ or $\gamma = 4/3$ stellar structures, which do not represent the range of stellar masses and stellar ages involved in TDEs. This work offers a significant improvement upon the Guillochon & Ramirez-Ruiz (2013) grid of simulations, and applies to any MS stellar mass and stellar age. The fitting formulae (see the Appendix) we find for key disruption quantities can be used to predict the disruption quantities for any star, provided one knows the $\rho_c/\rho$. Additionally, the STARS_library tool (Law-Smith et al. 2020, see Appendix) allows one to retrieve the $dM/dt$ for any stellar mass and age.

Our results are consistent with the hydrodynamical simulations of Goicovic et al. (2019), Gafton & Rosswog (2019) in the nonrelativistic limit, and Golightly et al. (2019b) for the impact parameters and stars they consider. The asymptotic decay power-law indices we find are broadly consistent with the analytic predictions of Coughlin & Nixon (2019) ($n_\infty = -9/4$ for partial disruptions, $n_\infty = -5/3$ for full disruptions), though our results indicate a more nuanced $\beta$ dependence (see Figure 13 and associated discussion). We note that we find $n_\infty \approx -2.2$ for partial disruptions, slightly larger than the $n_\infty = -2.25$ found by Coughlin & Nixon (2019).

Our results are broadly consistent with those of Ryu et al. (2020a, 2020b, 2020c, 2020d), though we note some differences. Rather than comparing $dM/dt$'s in detail, we note that the critical impact parameters we find for full disruption differ slightly. This is partly due to the fact that Ryu et al. (2020a, 2020b, 2020c, 2020d) study a single stellar age (MAMS), whereas we study two or three stellar ages for each star, so a direct comparison of results for a given stellar mass is misleading, as the stellar structures differ. See Figure 19 in the Appendix for a comparison. One disagreement is at the 15% level in our $\beta_{\text{crit}}$ = 0.9 result for the 0.3 $M_\odot$ star (which has an identical profile to a $\gamma = 5/3$ polytrope) for $M_{100} = 10^6 M_\odot$; Ryu et al. (2020a, 2020b, 2020c, 2020d) argue that this discrepancy is due to GR effects at $r_p = 22 r_g$. The difference is modest, but we note that our $\beta_{\text{crit}} = 0.9$ result has been independently found by other authors for polytropic stellar models, and the nonrelativistic regime is one in which our grid-based adaptive-mesh FLASH framework has also been independently verified with two other numerical methods (SPH and moving mesh) and resolutions by other authors (Guillochon & Ramirez-Ruiz 2013; Mainetti et al. 2017; Gafton & Rosswog 2019; Goicovic et al. 2019; Golightly et al. 2019b).

4.2. Caveats/Extensions

Our calculations are Newtonian, and thus do not capture the minority of encounters in which relativistic effects significantly alter the $dM/dt$. Cheng & Bogdanović (2014), Tejeda et al. (2017), and Gafton & Rosswog (2019) (for a review of TDEs in general relativity (GR) see also Stone et al. 2019) studied differences between the fallback rates from relativistic versus Newtonian tidal disruption simulations in detail, and found that for $r_p > 10 r_g$, the difference is $\lesssim 10\%$, and that the most significant corrections from relativistic effects are at pericenter distances smaller than this. They found that for $r_p > 10 r_g$ the error in a nonrelativistic tidal disruption simulation scales approximately as $v^2_p \sim r_g/r_p$ (though we emphasize that GR is a nonlinear theory, and order unity effects are possible for the rare extreme encounters in which GR dominates the dynamics). For

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Figure 12. $M_{\text{peak}} \times t_{\text{peak}}/M_\odot$ as a function of $\beta$ scaled with structural parameter $\alpha = (\rho_c/\rho)^{1/3}$.

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$n_\infty \approx -2.2$ (±0.4; Coughlin & Nixon 2019), while full and post-critical disruptions generally have $n_\infty \approx -5/3$, though the $\beta$-dependence is more nuanced than this. For example, $n_\infty$ peaks near the critical $\beta$ for full disruption, then falls slightly for post-critical disruptions. There is also a small region at small $\beta$ where there is a sharp transition from higher $n_\infty$ to $n_\infty \approx -2.2$—this was better captured in Guillochon & Ramirez-Ruiz (2013), who sampled $\beta$ more finely in this region.

In the bottom-left panel, one can see that the more centrally concentrated stars (which correspond roughly to the more massive stars) have $n_\infty \approx -2.2$ for progressively higher $\beta$'s. Aside from fitting individual TDEs, one can use this information with a statistical population of TDEs: given that the TDE rate depends inversely on $\beta$ to some power (see Section 4 for more detailed discussion), one can use $n_\infty$ as a probe of the stellar structure and more indirectly the stellar mass and age. More specifically, the lower-mass stars spend only a small portion of $\beta$ parameter space at $n_\infty \approx -2.2$, while the higher-mass stars spend much more of $\beta$ parameter space here, especially when weighted inversely by $\beta$ to some power. If the stellar mass function of TDEs is roughly flat for $M_* \lesssim M_\odot$ (Kochanek 2016a), and especially if mass segregation in galactic centers is important (see Section 4) then if an $n_\infty \approx -2.2$ is observed, it is more likely due to the partial disruption of a more massive star. Less massive stars are more likely to be full disruptions, and thus if $n_\infty \approx -5/3$ is observed, it is more likely due to the full disruption of a lower-mass star.

In the bottom-right panel, after rescaling the $x$-axis with $r_p/\rho$, the simulations reduce into a single relationship describing $n_\infty$ for any star and impact parameter. We provide a B-spline fit to this relation in the Appendix, so that one can obtain the $n_\infty$ for any stellar mass, stellar age, and impact parameter.

4. Conclusion

4.1. Comparison to Other Work

We compare with Guillochon & Ramirez-Ruiz (2013) throughout, but perhaps the most salient point is that, as we stress in this paper, every tidal disruption quantity that we have calculated depends primarily on the stellar structure, parameterized by $\rho_c/\rho$, and impact parameter $\beta$. Thus, the study of Guillochon & Ramirez-Ruiz (2013) is only directly applicable to

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Figure 13 and associated discussion. We note that we find $n_\infty \approx -2.2$ for partial disruptions, slightly larger than the $n_\infty = -2.25$ found by Coughlin & Nixon (2019).

Our results are broadly consistent with those of Ryu et al. (2020a, 2020b, 2020c, 2020d), though we note some differences. Rather than comparing $dM/dt$'s in detail, we note that the critical impact parameters we find for full disruption differ slightly. This is partly due to the fact that Ryu et al. (2020a, 2020b, 2020c, 2020d) study a single stellar age (MAMS), whereas we study two or three stellar ages for each star, so a direct comparison of results for a given stellar mass is misleading, as the stellar structures differ. See Figure 19 in the Appendix for a comparison. One disagreement is at the 15% level in our $\beta_{\text{crit}} = 0.9$ result for the 0.3 $M_\odot$ star (which has an identical profile to a $\gamma = 5/3$ polytrope) for $M_{100} = 10^6 M_\odot$; Ryu et al. (2020a, 2020b, 2020c, 2020d) argue that this discrepancy is due to GR effects at $r_p = 22 r_g$. The difference is modest, but we note that our $\beta_{\text{crit}} = 0.9$ result has been independently found by other authors for polytropic stellar models, and the nonrelativistic regime is one in which our grid-based adaptive-mesh FLASH framework has also been independently verified with two other numerical methods (SPH and moving mesh) and resolutions by other authors (Guillochon & Ramirez-Ruiz 2013; Mainetti et al. 2017; Gafton & Rosswog 2019; Goicovic et al. 2019; Golightly et al. 2019b).

4.2. Caveats/Extensions

Our calculations are Newtonian, and thus do not capture the minority of encounters in which relativistic effects significantly alter the $dM/dt$. Cheng & Bogdanović (2014), Tejeda et al. (2017), and Gafton & Rosswog (2019) (for a review of TDEs in general relativity (GR) see also Stone et al. 2019) studied differences between the fallback rates from relativistic versus Newtonian tidal disruption simulations in detail, and found that for $r_p > 10 r_g$, the difference is $\lesssim 10\%$, and that the most significant corrections from relativistic effects are at pericenter distances smaller than this. They found that for $r_p > 10 r_g$ the error in a nonrelativistic tidal disruption simulation scales approximately as $v^2_p \sim r_g/r_p$ (though we emphasize that GR is a nonlinear theory, and order unity effects are possible for the rare extreme encounters in which GR dominates the dynamics). For

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9 Though we note that Ryu et al. (2020d) argue for slightly larger effects due to the inclusion of GR.
encounters in which relativistic effects are significant, the decay power-law index remains approximately the same, and the most important differences are that the peak time is shifted to the right and the peak fallback rate is shifted down compared to the Newtonian case. The fraction of disruptions that require relativistic hydrodynamics in order to accurately model the mass fallback rate is \( \sim \max(1, 10r_g/r_t) \) in the pinhole regime of loss cone repopulation, and a step function in the diffusive regime: 

\[
\frac{1}{\alpha} \quad \text{if} \quad 10r_g < r_t, \quad \text{and} \quad \frac{1}{\alpha} \quad \text{if} \quad 10r_g > r_t.
\]

For a 1M_\odot, 1R_\odot, star, this is \( \approx 5\% \) for a 10^9M_\odot BH and \( \approx 20\% \) for a 10^7M_\odot BH. See Figure 20 in the Appendix for the range of applicability of Newtonian hydrodynamics simulations. Note that while relativistic effects may cause significant apsidal precession, the effect on the mass fallback rate to the BH is relatively small—and it is this quantity that tracks the luminosity evolution of TDEs so closely. While the library presented in this paper applies to the majority of TDEs, we plan to extend our setup to include relativistic encounters in future work.

Other extensions include studying orbits with different eccentricities, incorporating stellar spin (rather than performing new simulations with spinning stars, we could scale our \( dM/dt \) results with a parameterized stellar spin dependence based on the results of Golightly et al. 2019a), studying magnetic fields (we already have the framework for this, first studied in Guillochon & McCourt 2017), adding more stellar masses and ages to the grid, sampling more finely in \( \beta \) (though this will not make a significant difference because of our \( dM/dt \) interpolation), and various extensions and upgraded functionality of the STARS_library tool (Law-Smith et al. 2020). Finally, we plan to use the interpolated fallback-rate library as the new backbone for the publicly available TDE fitting software MOSFiT (Guillochon et al. 2018; Mockler et al. 2019).

We do not perform an analysis of TDE rates as a function of stellar mass and stellar age, but below is a brief discussion. In terms of broad demographics, Kochanek (2016a) calculated that for \( M_\mathrm{BH} \lesssim 10^9M_\odot \), the typical TDE is due to a 0.3M_\odot star, but that the mass function is relatively flat for \( M_* \lesssim M_c \). This study only considered effects due to the initial mass function (IMF) and did not include any dynamical interactions between stars, such as mass segregation in galactic centers, which segregates more massive stars closer to the BH and ejects less massive stars (e.g., Baumgardt et al. 2004; MacLeod et al. 2016)—this would cause the stellar mass function for TDEs to peak at higher masses than suggested by a convolution of the IMF and the luminosity function of TDEs (De Colle et al. 2012;
MacLeod et al. (2013). See also Figure 15 in MacLeod et al. (2012) for the fraction of stars at different evolutionary stages contributing to TDE flares. For $M_{\text{BH}} \lesssim 10^6 M_\odot$, MS stars make up $\gtrsim 85\%$. In more detail, MacLeod et al. (2012) find that the tidal disruption rate scales with the tidal radius of a given star as $\dot{n} \propto r_\text{t}^\alpha$, where $\alpha \approx 0.2$–0.5. Ignoring other considerations, the physical cross section increases with stellar age and leads to higher rates of tidal disruption for older MS stars. However, the $\Delta M$ lost at a given $\beta$ decreases as the star becomes more centrally concentrated, leading to fainter flares.

The dynamical mechanisms operating in the relevant galactic center and the associated most likely age for a star to be disrupted are also important. The two-body relaxation time, which is $\approx 2$ Gyr for a $10^4 M_\odot$ BH (Bar-Oz et al. 2013) (the MS lifetime of a $1.5 M_\odot$ star), gives a characteristic limit for the approximate mass and age of a star upon disruption. However, there are many other mechanisms that disrupt stars earlier. Stars can interact with one another coherently inside the sphere of influence of the BH, in contrast to two-body relaxation, leading to rapid angular momentum evolution (Rauch & Tremaine 1996); for example, this occurs for eccentric nuclear disks (Madigan et al. 2018). The presence of a supermassive black hole (SMBH) binary can also dramatically increase the TDE rate, driven by a combination of strong scatterings and Kozió–Lidov cycles between individual stars and the SMBH binary (e.g., Li et al. 2015). Other mechanisms that may disrupt stars over timescales $\ll 2$ Gyr include strong (non-diffusive) two-body scatterings, which make up a fraction of the TDE rate in any galaxy (Weissbein & Sari 2017), and radially anisotropic velocity fields or unusually dense star clusters, which may occur in post-starburst galaxies (Stone et al. 2018). Moreover, the fact that TDEs appear to be observed preferentially in rare post-starburst galaxies (Arcavi et al. 2014; French et al. 2016; Law-Smith et al. 2017b; Graur et al. 2018) means that many TDEs are sourced by unique stellar populations with particular dynamical histories.

4.3. Summary

We summarize the main points of this paper below.

1. We present a grid of tidal disruption simulations of stars with accurate stellar structures and chemical abundances, using MESA models as initial conditions to FLASH simulations with a Helmholtz EOS.

2. We interpolated this grid in 3D (stellar mass, stellar age, and impact parameter) to provide the STARS_library tool (see Appendix A), where one can request the $dM/dt$ for any stellar mass, stellar age, and impact parameter.

3. The quantities $\beta_{\text{crit}}$, $\Delta M$, $t_{\text{peak}}$, $M_{\text{peak}}$, and $n_{\infty}$ vary significantly with stellar mass and stellar age, but can be reduced to a single relationship that depends only on stellar structure, parameterized by $\rho_i/\bar{\rho}$, and impact parameter $\beta$ (see Figures 5, 6, 11, and 13). We provide fitting formulae for these quantities applicable to any MS star (see Figure 16 in the Appendix).

4. For the same $\Delta M$, the shape of the $dM/dt$ curve varies significantly with stellar mass, promising the potential determination of stellar properties from the TDE light curve alone (see Figure 10).

5. The critical impact parameter for full disruption increases with increasing central concentration, and scales approximately as $(\rho_i/\bar{\rho})^{2/3}$ for $\rho_i/\bar{\rho} \lesssim 500$ and $(\rho_i/\bar{\rho})^{1/3}$ for $\rho_i/\bar{\rho} \gtrsim 500$ (see Figure 6).

6. In general, more centrally concentrated stars have steeper $dM/dt$ rise slopes and shallower decay slopes (see Figures 8 and 9).

7. We show that the shape of $dM/dt$ depends only on the stellar density profile, and has little dependence on the internal EOS of the star (see Appendix G where we compare a $1 M_\odot$ and $10 M_\odot$ star with nearly identical density profiles), thus extending the range of applicability of our interpolated library and fitting formulae to any MS star.

We hope the community makes use of the STARS_library tool (see the Appendix), and look forward to incorporating this library as the new backbone of the MOSFiT fitting software for TDEs, or any other TDE fitting framework, allowing for more accurate determinations of both the nature of the disrupted star and the BH.

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Software: MESA (Paxton et al. 2011), FLASH (Fryxell et al. 2000), astropy (Astropy Collaboration et al. 2013, 2018), yt (Turk et al. 2011), NVIDIA IndeX (https://developer.nvidia.com/nvidia-index), STARS_library (Law-Smith et al. 2020).

Appendix A

Interpolated Fallback-rate Library

The STARS library of interpolated fallback rates, as well as up-to-date instructions for loading and using the library, is available on GitHub10 and version 1.0.0 is archived on Zenodo (Law-Smith et al. 2020). One can create custom interpolated grids and one can also query for any stellar mass, stellar age, and impact parameter.

The basic interpolation between $dM/dt$ is the same as in the MOSFiT software (Guillochon et al. 2018; Mockler et al. 2019). We first interpolate in $\beta$ for a given stellar mass and age. We then interpolate each of these $\beta$‘s in stellar mass, for a given fractional MS stellar age (e.g., $0.3 M_\odot$ ZAMS to $0.5 M_\odot$ ZAMS). We then interpolate in stellar age for a given stellar mass. In the query/retrieval mode, in order to retrieve a $dM/dt$ corresponding to a specific stellar mass, stellar age, and impact parameter that does not already exist in the interpolated library, we perform a quick series of interpolations on the nearest neighbor points in 3D space.

Figure 14 shows the STARS library interpolated $dM/dt$‘s for a small grid in stellar mass, stellar age, and impact parameter, for a single BH mass $M_{\text{BH}} = 10^6 M_\odot$. This grid has 10 interpolated $\beta$ points, 11 interpolated mass points, and five interpolated age points, with spacing in $\beta$ and stellar mass logarithmic, and spacing in stellar age linear.

10 STARS_library: github.com/jamielaw-smith/STARS_library.
Figure 14. STARS library interpolated $\frac{dM}{dt}$'s for a small grid of stellar mass, stellar age, and impact parameter, and at a single BH mass $M_{\text{BH}} = 10^6 M_\odot$. Axes labels are removed for clarity, but are the same as in other figures in this paper ($x$-axis is time, from 0 to 1000 days, $y$-axis is $\dot{M}$ from $10^{-4}$ to $10 M_\odot$ yr$^{-1}$, both log-scaled). Note that we provide more finely spaced interpolated grids for download (see text). The directory name, in format mXX_tYY, where XX is the stellar mass in $M_\odot$ and YY is the fractional MS age, is in the lower left of each panel.
Figure 15. STARS library interpolated $dM/dt$'s, all in one plot, for a single BH mass $M_{\text{BH}} = 10^6 M_\odot$. Left: absolute units. Right: normalized to peak time and peak fallback rate. We applied a fill between $dM/dt$'s in order to emphasize the tunable spacing of the 3D interpolation. Three $dM/dt$'s are overplotted in gray: a ZAMS 0.3$M_\odot$ star with $\beta = 0.6$ (dashed), a ZAMS 1$M_\odot$ star with $\beta = 4.2$ (solid), and a ZAMS 3$M_\odot$ star with $\beta = 2.0$ (dotted).

Figure 15 shows all of the STARS library interpolated $dM/dt$'s in the same plot, for a single BH mass $M_{\text{BH}} = 10^6 M_\odot$. The left panel is in absolute units, in order to emphasize the several orders of magnitude in $M$ and time covered by this grid even for a single BH mass—roughly six orders of magnitude in fallback rate and four orders of magnitude in time (we extend the $dM/dt$'s to later times than shown in this plot). The right panel is normalized to peak time and peak fallback rate, in order to emphasize the range of rise and decay slopes exhibited by the library. See Section 3 for more discussion of how these slopes depend on stellar structure. In general, more centrally concentrated stars have steeper rise slopes and shallower decay slopes. The more rounded behavior near peak is from post-critical-β disruptions; see e.g., the high-β behavior for a given star in Figure 7. We also overplot three $dM/dt$'s in gray: a ZAMS 0.3$M_\odot$ star with $\beta = 0.6$, a ZAMS 1$M_\odot$ star with $\beta = 4.2$, and a ZAMS 3$M_\odot$ star with $\beta = 2.0$.

Appendix B

Fits to Disruption Quantities

Figure 16 shows fitted B-splines and analytic formulæ for the mass lost from the star ($\Delta M/M_*$), time of peak mass fallback rate ($t_{\text{peak}}$), peak mass fallback rate ($M_{\text{peak}}$), and asymptotic decay power-law index ($n_\infty$). These fitting relations allow one to obtain the $\Delta M/M_*$, $t_{\text{peak}}$, $M_{\text{peak}}$, and $n_{\infty}$ for any stellar mass, stellar age, and impact parameter. One must first obtain the $\rho/\dot{\rho}$ value for this star, e.g., from Table 1, a MESA model directly, or a prepackaged grid of MESA models such as the MIST models (Choi et al. 2016).

The simple analytic fits are below. We were unable to fit the $n_{\infty}$ data with a simple formula, so the only option for this quantity is the B-spline.

$$\Delta M/M_* = 1.055 \tanh[(x + 0.04)^{20}] - 0.025$$  \hspace{1cm} (B1)

$$t_{\text{peak}} \text{[day]} = 2/x^{12} + 17x + 8$$  \hspace{1cm} (B2)

$$M_{\text{peak}}/(M_*/M_\odot) \text{[}M_\odot \text{yr}^{-1}] = 0.9 \sin(15x - 1.5)/x^{6} - 1.5x + 4.3$$  \hspace{1cm} (B3)

$$n_{\infty} = \text{(unable to find simple analytic fit; see B spline below)}$$  \hspace{1cm} (B4)

$$x = \exp[(\beta/\beta_{\text{crit}})^{\alpha} - 1], \hspace{0.5cm} \alpha = (\rho_{c}/\dot{\rho})^{-1/3}$$  \hspace{1cm} (B5)

$$\beta_{\text{crit}} \approx \begin{cases} 0.5(\rho_{c}/\dot{\rho})^{1/3}, \hspace{1cm} \rho_{c}/\dot{\rho} \lesssim 500 \\ 0.39(\rho_{c}/\dot{\rho})^{2/3}, \hspace{1cm} \rho_{c}/\dot{\rho} \gtrsim 500. \end{cases}$$  \hspace{1cm} (B6)

The range for $\Delta M/M_*$ is $x \in [0.8, 1.0]$ and for $t_{\text{peak}}$ and $M_{\text{peak}}$ is $x \in [0.8, 1.7]$. We note that the equations for $\beta_{\text{crit}}$ are only approximate.

Below we provide python code to read in and evaluate the B-spline fits to the disruption quantities $\Delta M/M_*$, $t_{\text{peak}}$, $M_{\text{peak}}$, and $n_{\infty}$. This code will reproduce the blue lines in Figure 16. Note that the order of all of the splines is 3. The knots and coefficients have been rounded to 3 decimals, which gives precision indistinguishable from the original fitted splines.

```
import numpy as np
from scipy.interpolate import splev

# deltam/mstar
knots = [0.804, 0.804, 0.804, 0.804, 0.804, 0.859, 0.881, 0.892, 0.903, 0.957, 1.001, 1.001, 1.001];
coeffs = [0.028, 0.068, 0.027, 0.159, 0.234, 0.416, 0.745, 0.97, 1.015, 0.0, 0.0, 0.0, 0.0, 0.0]

# tpeak
knots = [0.804, 0.804, 0.804, 0.804, 1.082, 1.638, 1.638, 1.638];
coeffs = [0.047, 0.067, 0.217, 0.617, 3.477, 1.714, 2.034, 2.024, 0.0, 0.0, 0.0, 0.0]

# mdotpeak/mstar
knots = [0.804, 0.804, 0.804, 0.804, 0.897, 0.943, 1.082, 1.638, 1.638, 1.638];
coeffs = [0.017, 0.127, 2.617, 3.477, 1.714, 2.034, 2.024, 0.0, 0.0, 0.0, 0.0]

# ninf
knots = [0.804, 0.804, 0.804, 0.804, 0.897, 0.943, 0.99, 1.036, 1.082, 1.175, 1.268, 1.314, 1.499, 1.638, 1.638, 1.638];
coeffs = [-2.079, -2.212, -2.19, -2.218, -1.333, -1.637, -1.652, -1.675, -1.73, -1.701, -1.671, -1.656, -1.695, 0.0, 0.0, 0.0, 0.0]

# evaluate B-splines
```
\[ x = \text{np.linspace}(\min(\text{knots}), \max(\text{knots}), 100); \]
\[ y = \text{splev}(x, [\text{knots}, \text{coeffs}, 3]) \]

Note that in order to use the B-splines to retrieve values for a specific star, one needs to use the scaled \( x \)-axis, and thus input the \( \rho_c/\bar{\rho} \) and \( \beta_{\text{crit}} \) values for the star. \( \rho_c/\bar{\rho} \) can be obtained as above and \( \beta_{\text{crit}} \) can be obtained from either Table 2 or our simple formula (Equation (B6)). Then one plugs these into

\[ x = \exp[(\beta/\beta_{\text{crit}})^{\alpha} - 1], \quad \alpha = (\rho_c/\bar{\rho})^{-1/3}. \]  

This \( x \) is the value to plug into the B-spline in order to retrieve the desired \( y \)-axis value.

**Appendix C**

**MESA Profiles Versus Polytropes**

Figure 17 shows MESA density profiles versus \( \gamma = 5/3 \) and \( \gamma = 4/3 \) polytropes, normalized to central density and stellar radius. One can see where a polytropic stellar model is sufficient and where we need to use a MESA stellar model as initial conditions for the FLASH simulations. For \( M_* = 0.1, 0.3M_\odot \), the profiles are nearly identical to \( \gamma = 5/3 \) polytropic stellar structures. For \( M_* = 0.5, 0.7M_\odot \) the profiles are in between \( \gamma = 5/3 \) \( \gamma = 4/3 \) polytropic stellar structures, and could in principle be simulated with a polytropic simulation using a \( \gamma \) that has been matched to this star. For \( M_* \gtrsim 1M_\odot \), as the star evolves off the ZAMS, the profile becomes more centrally concentrated than \( \gamma = 4/3 \) and thus cannot be simulated self-consistently with a polytropic stellar structure, as these are unstable for \( \gamma \lesssim 4/3 \). One requires a Helmholtz EOS in order to provide hydrodynamic support for these non-ZAMS stars. Note that the ZAMS \( M_* \gtrsim 1M_\odot \) stars are not uniformly well represented by \( \gamma = 4/3 \) polytropes—the only star that is quite well approximated by a \( \gamma = 4/3 \) polytrope is the \( 3M_\odot \) ZAMS star. Though we do not do FLASH simulations for intermediate ages between ZAMS and TAMS for the \( 1.5M_\odot, 3M_\odot \), and \( 10M_\odot \) stars in this work (we interpolate instead; see Appendix A), we show a MAMS profile at age \( t \approx t_{\text{TAMS}}/2 \) for these higher-mass stars (including for the \( 1M_\odot \) star, which we do simulate) for comparison. While the TAMS profiles are clearly much more centrally concentrated than a \( \gamma = 4/3 \) profile, the MAMS profiles are somewhat closer to the polytropic approximation, though still not a good fit.

One can also see that \( \rho_c/\bar{\rho} \) does not map 1:1 exactly to the stellar density profile, though it is a very good approximation. Consider the 10 Gyr \( 0.7M_\odot \) star \( (\rho_c/\bar{\rho} = 36) \) and the ZAMS \( 10M_\odot \) star \( (\rho_c/\bar{\rho} = 38) \). The \( 10M_\odot \) star has a slightly higher value of \( \rho_c/\bar{\rho} \) but its density profile is slightly less centrally concentrated than that of the \( 0.7M_\odot \) star.

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**Figure 16.** B-spline and analytic fits to disruption quantities \( \Delta M/M_* \), \( t_{\text{peak}} \), \( M_{\text{peak}} \), and \( n_\infty \). Note for the top-left panel, we extend \( \Delta M/M_* = 1 \) after full disruption (\( x \geq 1 \)).
Figure 17. MESA density profiles vs. $\gamma = 5/3$ and $\gamma = 4/3$ polytropes, normalized to central density and stellar radius. Each panel is a different mass. Blue is ZAMS, green is MAMS (shown for $M_* \geq 1M_\odot$), and orange is min(10 Gyr, TAMS).
Appendix D
Initial Profiles After Relaxation

Figure 18 shows the stellar density profiles after 5\(t_{\text{dyn}}\) of relaxation onto the hydrodynamical grid in FLASH, compared to initial profiles from MESA. To be clear, these two profiles match exactly at the start \(t = 0\) of the FLASH simulation, but here they are shown after 5\(t_{\text{dyn}}\). Our setup has been tested such that the density profile does not change significantly after \(\approx 100t_{\text{dyn}}\) of relaxation onto the hydrodynamical grid (without a BH present) in Guillochon et al. (2009).

As mentioned in Section 2, the most centrally concentrated stars, for which \(\rho_r/\rho \gtrsim 150\), have a higher initial resolution in FLASH of 512 cells across the diameter of the star. Stars with \(\rho_r/\rho \lesssim 150\) are initially resolved by 131 cells across their diameters.

Figure 18. Stellar density profiles after 5\(t_{\text{dyn}}\) of relaxation onto the hydrodynamical grid in FLASH (red), compared to initial profiles from MESA (black). These two profiles match exactly at the start of the simulation \(t = 0\).
Appendix E
Critical Impact Parameter and Pericenter Distance versus Stellar Mass

Figure 19 shows critical $\beta$ and critical pericenter distance over gravitational radius ($r_p/r_g$) as a function of stellar mass. See also Figure 6 for critical $\beta$ as a function of $\bar{\rho}/\bar{\rho}$, in which there is a simple power-law relationship. We find large ranges in critical $\beta$ and critical $r_p/r_g$, especially with stellar age for $M_* \gtrsim 0.8M_\odot$. At a fixed stellar mass, $\beta_{\text{crit}}$ can vary by a factor of $\approx 4$. We also compare to the results from Ryu et al. (2020a, 2020b, 2020c, 2020d). To avoid repetition, see Sections 2 and 4 for discussion of the differences between these works. We note that these authors studied only a single stellar age, so a direct comparison is often not possible. For physical pericenter distance, our average from 0.3 to $3M_\odot$ is $r_p/r_g \approx 20.4$. Our results are consistent with those of Ryu et al. (2020a, 2020b, 2020c, 2020d) for an $M_{\text{BH}} = 10^5M_\odot$ and for an $M_{\text{BH}} = 10^6M_\odot$ Ryu et al. (2020a, 2020b, 2020c, 2020d) find $r_p/r_g \sim 26.9$, averaging from 0.15 to $3M_\odot$.

Appendix F
Range of Applicability of Nonrelativistic TDE Hydrodynamics Simulations

Figure 20 shows regions where $r_p > 10r_g$ for a few stars ($0.1M_\odot, 1M_\odot, 10M_\odot$). This is where we expect nonrelativistic tidal disruption simulations to have $\lesssim 10\%$ error. For more detailed discussion of relativistic effects, see Section 4. Note that the BH masses of the host galaxies of observed TDEs, found independently through new observations and the $M-\sigma$ relationship (Wevers et al. 2017, 2019), lie in the range $3 \times 10^5M_\odot \leq M_{\text{BH}} \leq 2 \times 10^7M_\odot$ and peak at $10^6M_\odot$. The stellar masses obtained through fitting TDEs (Mockler et al. 2019) are almost all $M_* \lesssim 1M_\odot$.

Appendix G
Test of Stellar Structure versus EOS

In this paper we claim that stellar structure is the sole determinant of several tidal disruption quantities. To make this claim more robust we must first disentangle the effects of the EOS. As a test of the effects of stellar structure versus the EOS, we run simulations of two stars that have very similar density profiles but different energy support (radiative versus convective). We use a ZAMS $1M_\odot$ star (36 Myr) and a close-to-ZAMS $10M_\odot$ star (4 Myr). The $10M_\odot$ star is at central hydrogen fraction of $X = 0.65$, compared the ZAMS value of $X = 0.71$. The $1M_\odot$ and $10M_\odot$ star have very similar (but not
identical) normalized density profiles, and $\rho_\odot/\bar{\rho} = 42$ for the 1$M_\odot$ star and $\rho_\odot/\bar{\rho} = 45$ for the 10$M_\odot$ star. We chose to have the profiles match nearly exactly, rather than have the $\rho_\odot/\bar{\rho}$ values match. We ran one simulation at $\beta = 1.5$ for each, a "solid" partial disruption.

Figure 21 shows a comparison of the density profiles for these two stars. They are very different in absolute scales, but normalized to $\rho_\odot$ and $R_\odot$, the profiles are nearly indistinguishable.

Figure 22 shows the $dM/dt$‘s, absolute and scaled with $M_\odot$ and $R_\odot$. The 1$M_\odot$ star has $\Delta M/M = 0.581$ and the 10$M_\odot$ star has $\Delta M/M = 0.583$. The $dM/dt$‘s are very similar, which supports the argument advanced in this paper that the shape of the $dM/dt$ curve is determined by the stellar density profile, and allows us to provide the fitting formulae as a function of only $\rho_\odot/\bar{\rho}$ and $\beta$ for disruption quantities for any stellar mass and age within our simulation grid range. In fact, if, from this initial study of stellar structure versus EOS, we can conclude that these fitting formulae are robust for any star, then we can extend the scalings to stars with masses outside our grid range, as their different EOS’s will have only a small effect on the resulting disruption quantities.

The differences in shape between these two $dM/dt$’s could be due partially to the slightly different density profiles (note the small "notch" in the profile for the 10$M_\odot$ star in Figure 21) and values of $\rho_\odot/\bar{\rho}$ of these two stars. However—and this is the motivation for this test—it could be due to differences in the EOS of the two stars. In particular, the change in slope at $t \approx 100$ days may be due to different transitions between $\gamma = 5/3$ and $\gamma = 4/3$ pressure support in the $\rho$–$T$ plane for these two stars (e.g., see Figure 7 in Murguia-Berthier et al. 2017).

**Appendix H**

**MESA and FLASH Parameters**

In Tables 3 and 4 we list a few relevant parameters for the MESA and FLASH simulations. See Section 2 for explanations. MESA inlists are available upon request. We also turn on overshooting (not shown in the table) for $M_\star > 3M_\odot$ using the same parameter choices as in the MIST models (Choi et al. 2016).
Appendix I

Numerical Convergence

Figure 23 shows $dM/dt$'s for different resolution simulations in FLASH, demonstrating numerical convergence for the resolutions we choose in this paper: 131 cells across the diameter of the star for less centrally concentrated stars ($\overline{r}/\overline{\rho} \approx 150$) and 524 cells across the diameter of the star for highly centrally concentrated stars ($\overline{r}/\overline{\rho} \gtrsim 150$). We show results for a ZAMS 0.3 $M_\odot$ star (less centrally concentrated, $\overline{r}/\overline{\rho} = 6$), an MAMS 1$M_\odot$ star (moderately centrally concentrated, $\overline{r}/\overline{\rho} = 138$) and a TAMS 3$M_\odot$ star (highly centrally concentrated, $\overline{r}/\overline{\rho} = 1198$). We note that past simulations have had difficulty reaching convergence on the low-mass, “high-energy,” tightly bound tail of debris that returns first to the BH. This tail is of observational interest, and we achieve convergence here.

| Table 4 | A Few Relevant FLASH Parameters |
|---------|---------------------------------|
| Parameter | Value |
| $xmax$ | $1.e3$ |
| $tmax$ | $1.e2$ |
| $mpole_lmax$ | $20$ |
| $eos\_tolerance$ | $1.e-6$ |
| $smax\_l$ | $1.e7$ |
| $smax\_t$ | $1.e3$ |
| $smirho$ | $1.e-12$ |
| $smalp$ | $1.e0$ |
| $sim\_rho\_Ambient$ | $1.e-11$ |
| $sim\_p\_Ambient$ | $1.e1$ |
| $sim\_t\_Ambient$ | $1.e3$ |
| $order$ | $3$ |
| $slope\_Limiter$ | “hybrid” |
| $char\_Limiting$ | .true. |
| $use\_3d\_Full\_CTU$ | .true. |
| $use\_hybrid\_Order$ | .true. |
| $Riemann\_Solver$ | “HYBRID’’ |
| $shock\_Detect$ | .true. |

Note. $xmax$ is in units of $R_\odot$ and $tmax$ is in units of the stellar dynamical timescale $t_{dyn}$. 

Table 3

Relevant MESA Parameters

| Parameter | Value |
|-----------|-------|
| create\_pre\_main\_sequence\_model | .true. |
| new\_net\_name | 'mesa\_49\_net' |
| new\_rates\_prefix | 'a09' |
| initial\_zfracs | 6 AGSS09 |
| kappa\_lowT\_prefix | 'lowT\_fa05\_a09p' |
| kappa\_CO\_prefix | 'a09\_co' |
| initial\_z | $0.0142d0$ |
| initial\_y | $0.2703d0$ |
| L\_nuc\_div\_L\_zams\_limit | $0.999d0$ |
| mixing\_length\_alpha | $2$ |
| delta\_lg\_XH\_cntr\_hard\_limit | $0.00432d0$ |
| do\_element\_diffusion | .true. |
| xa\_central\_lower\_limit\_species(1) | ‘h1’ |
| xa\_central\_lower\_limit(1) | $0.001d0$ |

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Figure 23. $dM/dt$ is different resolution simulations in FLASH for a few different examples. Lines are labeled by the number of cells across the diameter of the star. Central concentration (parameterized by $\rho_c/\bar{\rho}$) is shown in the bottom left. Left: $0.3M_\odot$, 0 Gyr (a less centrally concentrated star; we use 131 cells here). Middle: $1.0M_\odot$, 4.8 Gyr (a moderately centrally concentrated star; we use 131 cells here). Right: $3.0M_\odot$, 0.3 Gyr (a highly centrally concentrated star; we use 524 cells here).
