Modelling massive star feedback with Monte Carlo radiation hydrodynamics: photoionization and radiation pressure in a turbulent cloud

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ABSTRACT

We simulate a self-gravitating, turbulent cloud of 1000 M☉ with photoionization and radiation pressure feedback from a 34 M☉ star. We use a detailed Monte Carlo radiative transfer scheme alongside the hydrodynamics to compute photoionization and thermal equilibrium with dust grains and multiple atomic species. Using these gas temperatures, dust temperatures, and ionization fractions, we produce self-consistent synthetic observations of line and continuum emission. We find that all material is dispersed from the (15.5 pc)³ grid within 1.6 Myr or 0.74 free-fall times. Mass exits with a peak flux of 2 × 10⁻³ M☉ yr⁻¹, showing efficient gas dispersal. The model without radiation pressure has a slight delay in the breakthrough of ionization, but overall its effects are negligible. 85 per cent of the volume, and 40 per cent of the mass, become ionized – dense filaments resist ionization and are swept up into spherical cores with pillars that point radially away from the ionizing star. We use free–free emission at 20 cm to estimate the production rate of ionizing photons. This is almost always underestimated: by a factor of a few at early stages, then by orders of magnitude as mass leaves the volume. We also test the ratio of dust continuum surface brightnesses at 450 and 850 μm to probe dust temperatures. This underestimates the actual temperature by more than a factor of 2 in areas of low column density or high line-of-sight temperature dispersion; the H II region cavity is particularly prone to this discrepancy. However, the probe is accurate in dense locations such as filaments.

Key words: hydrodynamics – radiative transfer – stars: massive – ISM: clouds – H II regions.

1 INTRODUCTION

Stars are formed in clusters within giant molecular clouds (GMCs), with observations in the Galaxy showing that the star formation efficiency (SFE), the fraction of the total mass in stars as opposed to gas, is a few per cent (Lada & Lada 2003). Numerical simulations of star formation tend to overestimate this proportion unless they invoke feedback mechanisms to drive down the SFE (Krumholz 2015), for example by introducing thermal feedback which can prevent fragmentation (Krumholz, Klein & McKee 2007; Bate 2009) or by removing reservoirs of gas that might otherwise condense into stars. This then effects the state of the stellar cluster once gas has been fully dispersed, as the exposed cluster may expand, lose stars, or become entirely unbound, depending on the SFE and dispersal time-scale (Lada, Margulis & Dearborn 1984).

Gas dispersal is thought to be driven by massive stars (of spectral type O or B and mass >8 M☉), since they emit ionizing radiation which heats gas to 10⁴ K, increasing thermal pressure and driving expansion on length-scales large enough to disrupt GMCs. The effect of ionizing feedback can be positive or negative with regards to the SFE. Elmegreen & Lada (1977) and Whitworth et al. (1994) showed using analytical arguments that the material collected together by shocks from expanding ionized gas can drive material into dense layers which then gravitationally fragments to form new stars. This is supported by numerical models of clouds irradiated by ionizing stars internally (Dale, Bonnell & Whitworth 2007a) as well as externally, with Dale, Clark & Bonnell (2007b) finding that feedback caused some stars to form earlier compared to control runs without feedback; furthermore, it caused the formation of new stars that would not have formed without feedback. On the other hand, the same simulation also displayed negative effects, with star formation being delayed for some objects, and overall the increase in SFE was small. Similarly, simulations by Walch et al. (2013) found that although triggering was effective on small time-scales, on large time-scales, the SFE was reduced due to the dispersal of gas. This was also supported by Geen, Soler & Hennebelle (2017),
whose models showed low SFEs consistent with observations of nearby clouds in the Galaxy.

Photoionization is not the only feedback process in play, however. As photons interact with gas and dust grains, they exert a radiation pressure which can change the morphology of H II regions and sculpt gas into shells (Draine 2011), or reduce the SFE by clearing cavities (Agertz et al. 2013). Massive stars also launch winds at high velocities (>1000 km s\(^{-1}\)), shocking gas to high temperatures (>10\(^5\) K; Krumholz 2015), and this may propagate out into the surrounding interstellar medium (ISM, Lopez et al. 2011). At the end of their lifetime, after a few Myr, massive stars explode as supernovae (SNe), injecting energy and momentum into the surroundings cleared out by feedback during the main sequence (Yorke et al. 1989; Rogers & Pittard 2013). The relative importance of these feedback processes is still not certain. In observational studies of the H II region 30 Doradus, Lopez et al. (2011) concluded that direct radiation pressure (from stellar photons) dominated over thermal pressure from ionized gas and wind-shocked gas, as well as indirect (dust-processed) radiation pressure. In 32 other H II regions, Lopez et al. (2014) observed that the ionized gas pressure was dominant, with two regions having a similar level of indirect radiation pressure, and all had significantly lower direct radiation pressure.

Analytical and numerical models provide a way to constrain the impact of feedback and much work has been done towards this at different length-scales and time-scales. Whitworth (1979) and Tenorio-Tagle (1979) modelled the dispersal of ionized gas via champagne flows in 1D, and this was built upon by Yorke et al. (1989) using 2D simulations with SNe exploding into the H II-region cavity. Models by Matzner (2002) concluded that expanding H II regions had a greater impact than stellar winds and SNe in driving turbulence within GMCs. Krumholz & Matzner (2009) determined that the impact of radiation pressure in the dynamics of H II regions increased with the number of massive stars and luminosity, and this requires high-surface density clouds (Fall, Krumholz & Matzner 2010). Peters et al. (2010) studied the growth of compact H II regions as protostars developed in a cluster, while Dale & Bonnell (2011) simulated larger scale GMCs, first with ionizing feedback, then later combining with stellar winds; the latter had less of an effect on the dynamics of large clouds, but in small clouds sculpted cavities through which ionizing radiation could leak (Dale et al. 2014). Rogers & Pittard (2013) simulated stellar winds and SNe and found that dense gas was largely resistant, with energy blowing out through low-density channels. More recently, Geen et al. (2015, 2016) combined photoionization with SNe, with the latter’s deposition of momentum depending on the number of ionizing sources pre-SN and the extent to which they had dispersed gas.

While numerical simulations such as these provide crucial information on the dynamics of star-forming regions, they must still be analysed in the same way that observers view real clouds to properly compare simulation with observation. It is therefore necessary to produce synthetic observations from the hydrodynamical models. This is a growing field, with examples such as Kurosawa et al. (2004), who used the Monte Carlo radiative transfer (MCRT) code TORUS to produce spectral energy distributions (SEDs) and Spitzer far-infrared observations of a completed smoothed particle hydrodynamics (SPH) simulation of accretion discs in a low-mass star-forming region. They used this data to verify the robustness of disc identification diagnostics. Haworth, Harries & Acreman (2012) synthesized metal forbidden-line images to calculate electron densities and temperatures of an H II region. Similarly, Koepferl et al. (2017a) used HYPERION (Robitaille 2011) and FLUXCOMPENSATOR (Koepferl & Robitaille 2017) to extract realistic observations of dust continuum from SPH models of massive star feedback by Dale et al. (2014). This allowed the testing of diagnostics used to calculate densities, temperatures, and star formation rates (Koepferl, Robitaille & Dale 2017b,c). Models by Dale, Ercolano & Bonnell (2012) were also post-processed by Hubber, Ercolano & Dale (2016) using MOCASSIN, another MCRT photoionization code, extracting emission lines from H, He, and metals.

The standard in modelling star-forming regions has so far been to carry out a radiation hydrodynamics (RHD) calculation with a simplified RT scheme, for example ray tracing to find the ionized Strömgren volume and setting the temperature inside to 10\(^5\) K (e.g. Walch et al. 2012; Dale et al. 2012). Snapshots from these models are then post-processed with a more detailed RT scheme to synthesize observations. However, since radiation and dynamics are physically intertwined every step of the way – temperature changes pressure which causes motion which sets a new condition for radiation, and so on – it would be more accurate to use the detailed microphysical prescription at every time-step of the RHD calculation. The resulting parameters, such as ionization states and temperatures, can then be fed into the synthetic observation processing without having to make post-hoc assumptions. This is what we set out to do in this paper. We describe our numerical methods in Section 2 and set out the initial conditions in Section 3. We present the results of the RHD model in Section 4, discuss them in Section 5, and show synthetic observations in Section 6. Finally, we summarize and conclude in Section 7.

2 NUMERICAL METHODS

We use the MCRT and hydrodynamics (HD) code TORUS as described by Harries (2015). Similar MCRT methods are used by Ercolano et al. (2003) and Wood, Mathis & Ercolano (2004), both of whom include detailed photoionization balance and dust physics. In TORUS, the MCRT is coupled with an HD module as described in Section 2.6, meaning an RHD simulation is carried out self-consistently.

The MCRT algorithm is based on Lucy (1999). At the beginning of a time-step of duration \(\Delta t\), the total stellar luminosity \(L\) is split into a total of \(N\) packets, each representing a bundle of photons of a particular frequency. Each packet has an energy

\[
\epsilon_i = \frac{L \Delta t}{N} \tag{1}
\]

with packets weighted with a factor \(w_i\) (whose sum is normalized to 1) depending on the frequency of the photons, such that a packet containing photons of a high frequency effectively has fewer photons and vice versa. Packets propagate through the grid with randomly sampled path lengths \(\ell\) between events representing absorption, scattering, or cell–boundary crossings. After absorption events, packets are re-emitted with a frequency sampled from a probability density function (PDF) constructed from the cell emissivities, and are appropriately reweighted with a new \(w_i\); thus TORUS uses a polychromatic treatment of both the stellar and diffuse radiation fields. If each path takes a time \(\delta t = \ell/c\), then for a total time-step \(\Delta t\), the path contributes an energy \(\epsilon_i \delta t \Delta t = \epsilon_i \ell/c \Delta t\) to the radiation field. The total energy density is then

\[
d\rho_{\nu} = \frac{1}{c \Delta t V} \sum \epsilon \ell \tag{2}
\]
where the sum is over events, and this is proportional to the mean intensity \(J_v\) via

\[
d\omega_v = \frac{4\pi J_v}{c} \, dv
\]

This allows the path length algorithm to be used for applications of radiative and photoionization equilibrium.

### 2.1 Photoionization balance

We use the photoionization algorithm described by Haworth & Harries (2012), which is similar to Ercolano et al. (2003). Equating the rates of photoionization and recombination gives the ratio of number densities \(n\) of successive ionization states \(i\) of species \(X\) (Osterbrock & Ferland 2006):

\[
\frac{n(X^{i+1})}{n(X^i)} = \frac{1}{n_e \alpha(X^i, T)} \int_{v_i}^\infty a_v(X^i) 4\pi J_v \, dv
\]

where \(n_e\) is the electron number density, \(\alpha\) is the total recombination coefficient to all energy levels, \(a_v\) is the absorption cross-section, \(\nu_i\) is the ionization frequency for species \(X^i\). Making use of equations (2) and (3), the mean intensity \(J_v\) can be replaced with MC estimators to give

\[
\frac{n(X^{i+1})}{n(X^i)} = \frac{1}{n_e \alpha(X^i)} \frac{1}{V \Delta t} \sum_{v_i} \frac{a_v(X^i) \epsilon \ell}{h \nu_v}
\]

Haworth et al. (2015) investigated the relative importance of different microphysics on the expansion of \(\text{HII}\) regions. As Lyman continuum photons are absorbed by dust grains and re-emitted in the infrared, Haworth et al. concluded that dust is important in the early stages of the evolution, as the number \(N_{d,0}\), that goes into ionizing the gas is reduced; the effect is to reduce the Strömgren radius \(r_S\) by about 10 per cent, since \(r_S \propto N_{d,0}^{-1/3}\). Haworth et al. also tested the sensitivity to temperature of the \(\text{HII}\) region expansion. By treating the radiation field polychromatically, including the contribution from the diffuse gas as well as stars, and also including cooling from helium and metals, temperatures are reduced such that the ionization radius at later times is reduced by about 10 per cent compared to models which only consider hydrogen and a monochromatic, on-the-spot treatment. Therefore, we use the full treatment in these cluster simulations.

### 2.2 Thermal balance

Thermal balance is calculated in the same way as Haworth et al. (2015). We calculate separate temperatures for the gas and dust in our models, which are only coupled by a heat-exchange rate to account for collisions.

Dust heating is calculated assuming radiative equilibrium using the Lucy (1999) algorithm. The rates at which gas absorbs and emits radiative energy are, respectively,

\[
\dot{A} = 4\pi \int_0^{\infty} k_v J_v \, dv
\]

\[
\dot{E} = 4\pi \int_0^{\infty} k_v B_v(T_d) \, dv
\]

where \(k_v\) is the opacity per unit length. Equation (6) can be rewritten in terms of calculable MC estimators,

\[
\dot{A} = \frac{1}{\Delta t V} \sum_{v_i} k_v \epsilon \ell
\]

\[
\dot{E} = \frac{1}{\Delta t V} \sum_{v_i} u \hat{u} \cdot k_v \hat{u} \cdot B_v(T_d)
\]

### Table 1. Total abundance of each element and the ionized states included in the simulation.

<table>
<thead>
<tr>
<th>Element</th>
<th>(\log_{10}(\text{abundance})) (rel. to H)</th>
<th>Ionized states</th>
</tr>
</thead>
<tbody>
<tr>
<td>Hydrogen</td>
<td>0</td>
<td>I–II</td>
</tr>
<tr>
<td>Helium</td>
<td>−1</td>
<td>I–III</td>
</tr>
<tr>
<td>Carbon</td>
<td>−3.66</td>
<td>I–IV</td>
</tr>
<tr>
<td>Nitrogen</td>
<td>−4.40</td>
<td>I–III</td>
</tr>
<tr>
<td>Oxygen</td>
<td>−3.48</td>
<td>I–III</td>
</tr>
<tr>
<td>Neon</td>
<td>−4.30</td>
<td>I–III</td>
</tr>
<tr>
<td>Sulphur</td>
<td>−5.05</td>
<td>I–IV</td>
</tr>
</tbody>
</table>

The emission rate can be simplified using the Planck mean opacity, \(k_\nu\), which leads to

\[
E = 4\pi k_\nu B(T_d) = 4\pi k_\nu \frac{\sigma T_d^4}{\pi}
\]

where \(B\) is the frequency-integrated Planck function and \(\sigma\) is the Stefan–Boltzmann constant. Setting this equal to the absorption rate gives the dust temperature

\[
T_d = \left( \frac{A}{4\pi k_\nu} \right)^{1/4}
\]

To find the gas temperature of each cell, we calculate the solution which gives the heating rate equal to the total cooling rate. Gas is heated via ionization of neutral H and He. For species \(X^i\) at ionization state \(i\), the heating rate is given by

\[
G(X^i) = n(X^i) \int_{v_i}^\infty a_v(X^i) \frac{4\pi J_v}{h \nu_v} (h \nu_v - h \nu_i) \, dv
\]

\[
= n(X^i) \frac{1}{V \Delta t} \sum_{v_i} \frac{a_v(X^i) \epsilon \ell}{h \nu_v} (h \nu_v - h \nu_i)
\]

Sources of gas cooling are free–free radiation, recombination lines from H and He, and collisionally excited forbidden lines from metals; Table 1 shows the species included in our model.

The collisional gas-dust heat exchange rate per unit volume is taken from Hollenbach & McKee (1979),

\[
\Gamma_{\text{gas-dust}} = 2 f n_d n_g \sigma_d v_p (T_g - T_d)
\]

where \(n_d, \sigma_d,\) and \(T_d\) are the number density, cross-section, and temperature of dust grains, \(v_p\) is the thermal speed of protons at the gas temperature \(T_g\), and \(f\) is a factor which depends on the ionization state and gas temperature.

### 2.3 Radiation pressure

We use the momentum-transfer scheme described by Harries (2015) to calculate the radiation pressure in each cell. As photon packets interact with a cell, they exchange momenta such that the net change in momentum between the packet leaving and entering the cell is

\[
\Delta p = \frac{\epsilon_i}{c} \Delta \hat{u}
\]

where \(\hat{u}\) is the unit vector in the direction of travel. The radiative force per unit volume is then

\[
\begin{aligned}
\vec{f}_{\text{rad}} &= \frac{\Delta p}{\Delta t V} \\
&= \frac{\Delta p}{\Delta t V}
\end{aligned}
\]

which is added on to the momentum equation during the HD step.
2.4 FUV interstellar radiation field
The far-ultraviolet (FUV) flux $G_0$ in units of the Habing flux (Habing 1968) is calculated in all grid cells at all time-steps using

$$G_0 = \int_{\lambda=2400\AA}^{912\AA} 4\pi f_\lambda d\nu = \frac{1}{\Delta t V} \sum_{\lambda=2400\AA}^{912\AA} \epsilon f$$

(15)

(Osterbrock & Ferland 2006).

2.5 MC estimator smoothing
In order to increase the efficiency of the MC estimators for the radiation field, we use a scheme where each of the previous estimates are weighted according to how many time-steps ago they occurred and then averaged. The weighting for each estimate of the radiation field is given by

$$u_k = \exp \left( \frac{k \Delta t}{t_{\text{rad}}} \right)$$

(16)

where $k$ is the number of time-steps ago the estimate was made, $\Delta t$ is the time-step of the simulation, and $t_{\text{rad}}$ is the radiation time-scale. For a sufficiently large number of previous estimates, the total of the weights can be approximated to the infinite sum

$$\sum_{k=0}^{\infty} e^{-ak} = \frac{e^a}{e^a - 1}$$

(17)

$$\alpha = \frac{\Delta t}{t_{\text{rad}}}$$

(18)

$$f_{\text{sum}} = \sum_{k=0}^{\infty} (f_k e^{-ak})$$

(19)

Using this formulation of the weights allows us to retain all the information for all the previous radiation history as a single value (equation 19). In order to calculate the weighted radiation value for the next time-step from the instant estimate of that time-step and the weighted sum from the previous time-step, we can use the fact that all the weights from the previous time-step are simply multiplied by $e^{-a}$ to give the weights for the next time-step. This allows us to calculate the new-weighted radiation value using

$$f_{\text{weighted}} = (f_n + e^{-a} f_{\text{sum}}) \frac{e^a - 1}{e^a}$$

(20)

Once this has been done $f_{\text{sum}}$ is set to the new value of $f_n + e^{-a} f_{\text{sum}}$ for the next time-step. For the value of $f_{\text{sum}}$ at $t = 0$, we assume the radiation field has been static for a long time so that $f_{\text{sum}} = f_0 \frac{e^{-ak}}{1 - e^{-ak}}$.

This method gives an improved estimate of the radiation field by drawing on more information at the cost of introducing some time lag into the radiation field as it changes.

2.6 Hydrodynamics
A hydrodynamics step takes place after each radiation step. TURB solves the Eulerian equations of mass conservation and momentum conservation,

$$\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho u) = 0$$

(21)

using a finite-volume, explicit-differencing method, assuming the hydrodynamics evolves isothermally. The equations are solved using an operator splitting method. A V-cycling multigrid method is used to solve Poisson’s equation,

$$\nabla^2 \phi = 4\pi G \rho$$

(23)

with Dirichlet boundary conditions based on a multipole expansion. Mass is allowed to flow out through the boundaries but not in.

2.7 Stars
Stars are represented by moving Lagrangian sink particles as implemented by Harries (2015), based on Federrath et al. (2010). This implementation was used by Harries, Douglas & Ali (2017) to model the formation of a single massive star on sub-parsec scales. In our simulation, since we do not resolve down to these scales, we do not initiate sink accretion, but we still include the gravitational forces. Stars begin on the zero-age main sequence and follow stellar evolution tracks by Schaller et al. (1992) between 0.8 and 120 $M_\odot$. We use the tracks of two masses, $M_1$, $M_2$, such that the initial stellar mass lies between $M_1 < M_{\text{ini}} < M_2$. We interpolate to find the new mass at the current age in the $M_1$ track, repeat for the $M_2$ track, then interpolate between the two resulting masses, yielding the final new mass. We follow the same procedure for the luminosity, effective temperature, and radius. Spectra of O stars follow the OSTAR2002 grid of models calculated using TLUSTY by Lanz & Hubeny (2003), while later type stars follow the models of Kurucz (1991).

3 INITIAL CONDITIONS
We carry out our simulations on a 3D grid with a uniform resolution of 256$^3$. This resolution is chosen such that the point of complete gas dispersal is reached within a reasonable computing wall time; we have also tested models at lower resolution and results converge at 256$^3$. The initial condition is a spherical cloud with a uniform-density inner core extending to half the sphere radius, with the outer half going as $r^{-1.5}$. The density outside the sphere is 1 per cent of the density at the sphere edge. These conditions are similar to other cluster simulations such as Krumholz, Klein & McKee (2011) and Howard, Pudritz & Harris (2016). The sphere has a total mass $M_\text{cl} = 1000 M_\odot$, radius $R = 2.66$ pc, and mean surface density $\Sigma = 0.01$ g cm$^{-2}$. The size of the grid is approximately six times the sphere radius at 15.5 pc, giving a resolution of 0.06 pc per cell. We impose the same random Gaussian turbulent velocity field as Bate, Bonnell & Bromm (2002), with a power spectrum $P(k) \propto k^{-4}$ for wavenumber $k$, such that the kinetic energy equals the gravitational potential energy, i.e. the virial parameter $\alpha_{\text{vir}} = 2 E_{\text{grav}} / E_{\text{kin}} = 2$.

We evolve the clouds under gravity and turbulence without stars up to 0.75 $(t_{\text{ff}})$, where $(t_{\text{ff}}) = 2.17$ Myr is the average free-fall time associated with a sphere of uniform density $\rho = 3 M_\odot / 4\pi R_\text{ff}^3$; this is when Krumholz et al. (2011) see an SFE of 10 per cent. At this time, we randomly sample stars from a Salpeter (1955) initial mass function such that the cumulative stellar mass is 10 per cent of the cloud mass ($100 M_\odot$) and at least one massive star is present. The most massive star ($33.7 M_\odot$) is placed in the cloud’s most massive clump. The other 28 stars (the next most massive being $11 M_\odot$) are placed according to a probability density function assuming a star formation rate $\dot{M}(r)$ at some position $r$; that is,
Table 2. Initial parameters of the massive star.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mass</td>
<td>$33.7 , M_{\odot}$</td>
</tr>
<tr>
<td>Luminosity</td>
<td>$1.49 \times 10^5 , L_{\odot}$</td>
</tr>
<tr>
<td>Radius</td>
<td>$7.59 , R_{\odot}$</td>
</tr>
<tr>
<td>Effective temperature</td>
<td>41 189 K</td>
</tr>
<tr>
<td>Ionizing flux ($h\nu \geq 13.6 , eV$)</td>
<td>$7.36 \times 10^{48} , s^{-1}$</td>
</tr>
</tbody>
</table>

Figure 1. Positions of stars at the onset of feedback, with stellar mass in colour scale, overlaid on column density in grey scale (both are logarithmic). The most massive star is $33.7 \, M_{\odot}$ in red. The second highest is $11.3 \, M_{\odot}$. The third is $5.7 \, M_{\odot}$. The least massive is $0.82 \, M_{\odot}$.

The initial radius, luminosity, effective temperature, and ionizing photon production rate of the massive star are listed in Table 2. The initial distribution of stars is shown in Fig. 1 overlaid on top of column density. At this point, we switch on the radiation field and evolve the simulation until all the mass leaves the volume.

Elemental abundances are listed in Table 1, using the same values as Haworth et al. (2015). We include the first few ionized states of each element, with ionization fractions calculated using the photoionization equilibrium equation (5). The total abundance of each element remains constant. We also include silicate dust grains using properties from Draine & Lee (1984) – Fig. 2 shows the dust absorption, scattering, and total opacities as a function of wavelength. We use a constant dust-to-gas ratio of 0.01 and follow a standard ISM power-law density distribution (Mathis, Rumpl & Nordsieck 1977)

$$n(a) \propto a^{-q}$$

(24)

using grain sizes $a$ between 0.1 and 1 μm and a power-law index $q$ of 3.5, giving a median grain size of 0.12 μm.

Figure 2. Dust opacity per unit mass as a function of wavelength showing scattering, absorption, and total opacities.

4 RESULTS

In this section, we present the results from a model with both photoionization and radiation pressure feedback, along with a model with just photoionization [i.e. $f_{rad}$ in equation (22) is set to zero].

4.1 Bulk grid properties

In Fig. 3, we plot as a function of time the total mass on grid, mass flux off the grid, maximum density, ionized mass and mass fraction, and ionized volume fraction. Fig. 4 shows column density at 0.2 Myr snapshots.

The peak mass flux off the grid is $2 \times 10^{-3} \, M_{\odot} \, yr^{-1}$, reached 0.6 Myr after initiating feedback – this is the ionized sound-crossing time from the centre of the grid to the boundary. From the onset of feedback ($t = 0$) to 0.6 Myr, the mass flow is steady, with low-density ionized gas streaming through the channels carved out of the high-density filaments by the expanding ionization front. The envelope of diffuse gas outside of the initial cloud is also pushed outwards by the expanding gas. The right half of the cluster (in Fig. 4) is dispersed more quickly than the left half, which contains higher density structures and is therefore more resistant to destruction and dispersal. They remain on the grid and close to their initial positions, but the ionization front creates holes in low-density areas, and curves around high-density areas, creating clumps with tails pointing away from the ionizing star. These objects move radially away from the massive star due to the rocket effect caused by photoevaporation along their star-side edges (as in e.g. Bertoldi & McKee 1990; Mellema et al. 1998). They approach the edge of the volume over the course of the simulation with an average speed $\approx 6 \, km \, s^{-1}$. After 0.6 Myr, the overall mass flux begins to decrease but contains spikes corresponding to the removal of the clumps. The size of the spikes grows with time, as the densest clumps are the last to leave the grid. By about 1.6 Myr, or 0.74 $\langle t_{ff} \rangle$, all the mass has left the (15.5 pc)$^3$ volume.

The third panel of Fig. 3 shows the maximum density in g cm$^{-3}$ as a function of time. This peaks at just under $4 \times 10^{-19} \, g \, cm^{-3}$ at 0.6 Myr. Between 0.2 and 0.6 Myr are when the densities become highest, as the expansion of the H$\alpha$ region drives material together. In the first 0.2 Myr, as gas gets ionized, the dense core containing the massive star expands spherically outwards, colliding with another
Figure 3. Bulk grid properties as a function of time, showing total mass, mass flux off the grid, maximum mass volume density, ionized mass, ionized mass fraction, and ionized volume fraction. The blue line is the model with ionization and radiation pressure; and the green line is only ionization. \( t = 0 \) corresponds to the onset of feedback.

Figure 4. Column density integrated along the \( z \)-axis at 0.2 Myr intervals in the combined feedback model. Each frame is 15.5 pc a side. The 34 M\(_{\odot}\) star is indicated with a point near the centre of each frame. \( t = 0 \) corresponds to the onset of feedback.
set of dense filaments nearby \((\approx 10^{-19} \, \text{g cm}^{-3})\); at \((x, y) \approx (−1, 0) \, \text{pc}\) in Fig. 4, which shows column density). Between 0.2 and 0.4 Myr, the outflung material sweeps across the filament, with the densest areas remaining somewhat stationary, while the lower density gas is carried along with the flow. During this process, the filament is compressed and material that is initially perpendicular to the expanding shell is broken up and carved into pillars oriented parallel to the flow, ending up as tails behind spherical cores (pointing radially away from the ionizing star). Compression of the filament causes the maximum density to increase, reaching its highest value of \(4 \times 10^{-19} \, \text{g cm}^{-3}\) at 0.6 Myr. Once the expanding material has passed through and the pillars are formed, the densest cores are more exposed to the stellar radiation field and there is less collisional compression – photoevaporation removes material from the clumps, parts of the pillars break off into separate chunks, and the maximum density falls.

The fourth and fifth panels of Fig. 3 show the total ionized mass and the ionized mass fraction, respectively. The highest value of ionized mass is 440 \(M_\odot\) at 0.5 Myr (36 per cent of the total mass). The peak ionized mass fraction is reached 0.1 Myr later, still just under 40 per cent of the total mass. This is despite 85 per cent of the volume being ionized (sixth panel of Fig. 3), showing that most of the mass remains in small, dense clumps which resist ionization.

The photoionization-only model is mostly similar to the model with both photoionization and radiation pressure, but there are a few minor differences. The bulk effect of feedback is delayed in the photoionization-only model, with the peak in the ionized mass fraction being reached 0.03 Myr later. The later breakout of ionization results in a slightly different distribution of gas, as it has had more time to evolve under gravity and turbulence, so the total amount of gas being ionized is affected (very marginally) – the peak ionized mass fraction is about 2 per cent higher. The removal of gas from the grid also occurs with the same delay. Overall, the differences are negligible.

### 4.2 Morphology

Snapshots of column density in Fig. 4 show how the destruction of the cloud proceeds via the expansion of ionized gas over the course of 1.6 Myr in the model containing both photoionization and radiation pressure. Fig. 5 shows a 2D slice of mass density and velocity vectors at 0.1 Myr. Most of the \(\text{H} II\) region is confined at 0.1 Myr by high-density filaments, but a champagne flow breaks out through the low-density region on the edge of the cloud on the opposite side. At the boundaries of the \(\text{H} II\) region, the gas travels outwards at the ionized sound speed (approximately 12 km s\(^{-1}\)), while the photoevaporation of gas on the inside of the boundary leads to outflows moving inwards and then out through the champagne flow at velocities of 20–30 km s\(^{-1}\), with a few cells around 40 km s\(^{-1}\).

The \(\text{H} II\) region is able to break through some parts of the filament [e.g. near \((x, y) = (0.8, 1.4) \, \text{pc}\) in Fig. 5], while curving around nodes such as \((x, y) = (0.1, 0.1) \, \text{pc}\) which resist photoionization and are carved into globules and pillars by the expanding ionization front. The densest clumps shield material that is downwind of the ionizing source as seen in Fig. 6, which shows the ionization fraction of hydrogen at 0.6 Myr. Some of the shielding is only partial, for example at \((-4, 1.5) \, \text{pc}\) of that frame, as the diffuse radiation field ionizes gas behind the clump but to a lesser degree. This highlights the importance of including the diffuse field in RHD models.

Fig. 7 shows column density histograms for the combined feedback model at 0.2 Myr intervals. The spike at \(t = 0\) (the onset of feedback) at \(\Sigma \approx 3 \times 10^{-4} \, \text{g cm}^{-2}\) corresponds to the stationary gas outside around the gravoturbulent cloud, which is only perturbed when the star photoionizes it or cloud material expands into it, at which point the surrounding density increases. Overall, as the simulation evolves, the PDF shifts towards lower densities, as a result of the \(\text{H} II\) region expansion and the increase in low-density, ionized material. The early high-density hump at \(\approx 10^{-2} \, \text{g cm}^{-2}\), produced during the initial starless collapse phase, is flattened out in the first 0.4 Myr. Although higher densities are achieved up to the same period, they are not long-lived, as the maximum density is reduced after another 0.2 Myr.

Figure 5. Zoomed-in slice in the \(xy\)-plane at 0.1 Myr showing mass density in grey scale between \(10^{-23}\) and \(9.4 \times 10^{-20} \, \text{g cm}^{-3}\); vectors with size and colour corresponding to velocity between 0 and 60 km s\(^{-1}\); and a contour where the ionization fraction is 0.9. The 33.7 \(M_\odot\) star is at position \((x, y) = (0.6, 0.1) \, \text{pc}\). This model includes ionization and radiation pressure.

Figure 6. 2D slice in the \(xy\)-plane at 0.6 Myr showing the hydrogen ionization fraction. The linear grey scale is the fraction between 0.9 and 1 (fully ionized gas). The logarithmic colour scale is the fraction between 0.1 and 0.9 (essentially fully neutral to partially ionized). The densest clumps and pillars are neutral, but have partially ionized edges. The interior is fully ionized.
4.3 Electron temperature and density

In order to gauge the temperature of the ionized gas, we calculate weighted averages over the volume using

$$ T_0 = \frac{\int w T dV}{\int w dV} = \frac{\sum_i w_i T_i \Delta V}{\sum_i w_i \Delta V} $$

(25)

where $T_i$ is the temperature of cell $i$ with volume $\Delta V$, and we consider two different weights $w$: (i) $w = n_e (n_{\text{HII}} + n_{\text{HeII}})$ as per Rubin (1968), which is well approximated by $w \approx n_e^2$; (ii) mass $w = \rho \Delta V$ if hydrogen in the cell is more than 90 per cent ionized, or $w = 0$ if less (green). Solid lines are the model with ionization and radiation pressure; and dashed lines are the model with just ionization.

The mass-weighted average temperature is highest at 0.2 Myr, having a value 9300 K, whilst the ionization front is still largely contained inside the cloud, but then decreases to 8000 K over the next 1.4 Myr. The $n_e^2$-average is 9000 K over the whole duration. The standard deviation is about 10 per cent for the $n_e^2$ average and is steady until about 1 Myr, at which point the deviation rises, with greater fluctuation, towards 14 per cent. The mass-weighted average also has a standard deviation around 10 per cent, but after 0.4 Myr, this drops to 6 per cent and is much more steady than the electron density-weighted average. This is due to the high-mass globules and filaments which are neutral (and hence do not contribute to either average), but whose edges are partially ionized (so they do contribute to the electron density-weighted average, but not the mass-weighted average). This is visualized in Fig. 6 where the grey scale denotes cells which are more than 90 per cent ionized, and the colour scale shows cells which are less. Since the interaction of the ionization front and the dense, neutral gas changes relatively quickly over time, the spatial extent and degree of partial ionization similarly changes, giving rise to the fluctuation in the $n_e^2$ average.

Fig. 9 shows the (unweighted) volume-average electron density in gas which is more than 90 per cent ionized. This reaches a maximum of 30 cm$^{-3}$ at $\approx$0.1 Myr before dropping down to 8 cm$^{-3}$ by 0.2 Myr. It remains at this value until 0.5 Myr. As the gas flows off the grid over the next Myr, the density decreases once again, reaching 0.2 cm$^{-3}$ by the end of the simulation.

4.4 FUV interstellar radiation field

Fig. 10 shows a representative 2D slice of $G_0$, the FUV flux (equation 15), in the plane of the 34 M$_\odot$ star, showing the variation with distance at $t = 0.4$ Myr. The flux decreases as $r^{-2}e^{-\tau}$ due to geometrical dilution and the optical depth $\tau$, as in Bisbas et al. (2015), with shadowed cones caused by dense, dusty clumps which shield the downstream material from the stellar radiation field. The maximum value is of the order 10$^6$ in the cell containing the star, with values of order 10$^5$ in adjacent cells. In Fig. 11, we plot as function of time the average $G_0$ inside a radius of 2.5 cells (0.15 pc) around the star, weighted by mass. Averaging over time, this is $(2.0 \pm 0.3) \times 10^5$ in units of the Habing flux. Fluctuations arise from the star moving between cells which have different densities and therefore different optical depths. The inner Orion nebula is observed to have values around $4 \times 10^5$ (Osterbrock & Ferland 2006), which is comparable in magnitude to the region we model here.
Figure 10. Interstellar FUV flux $G_0$ in units of the Habing flux $(1.63 \times 10^{-3} \text{ erg s}^{-1} \text{ cm}^{-2})$. This is a slice in the $xz$-plane at 0.4 Myr, taken through the position of the most massive star (pink circle, $34 \, M_\odot$). Stars are plotted as 3D surfaces with a radius of 2.5 grid cells, therefore intersections of the slice with star surfaces result in circles (rainbow colour scale, with the size corresponding to the proximity to the slice). Grey-scale contours show mass volume density at levels of $10^{-19}$, $10^{-20}$, $10^{-21}$, and $10^{-22} \text{ g cm}^{-3}$.

Figure 11. Interstellar FUV flux $G_0$ in units of the Habing flux $(1.63 \times 10^{-3} \text{ erg s}^{-1} \text{ cm}^{-2})$, at the location of the most massive star, as a function of time. This is the mass-weighted average within a radius of 2.5 grid cells (0.15 pc) around the star. The horizontal dashed black line shows the time-average flux and the dotted lines show the standard deviation. We include a similar plot for a $0.88 \, M_\odot$ star. This is shown in Fig. 12 alongside the time-varying distance to the massive star. Between 0 and 0.5 Myr, $G_0$ rises steadily to a value of $2 \times 10^3$, before jumping up by more than a factor of 3.5 within $\approx 50$ kyr. This occurs because it is located inside a pillar which is being photoevaporated and pushed away from the ionizing source, decreasing the column density between the two stars; once it becomes completely exposed to the radiation field of the brightest source, the flux rapidly rises. After this point, $G_0$ remains relatively level up to the end of the simulation as the intervening material is diffuse, with a slight increase as the stars approach each other. We save a detailed investigation of the fluxes around the other 27 stars for a future study.

5 DISCUSSION OF THE RHD MODEL

5.1 Dynamics

The differences between the model with both ionization and radiation pressure, and the model with just ionization, are negligible, leading to the conclusion that photoionization is a more important process than radiation pressure for the dispersal of gas in the conditions presented here. According to Draine (2011), the latter process is expected to have a more significant role in higher density clouds, e.g. $n > 100 \text{ cm}^{-3}$ with a star having the same Lyman continuum photon flux as our model ($\approx 10^{49} \text{ s}^{-1}$).

Gas dispersal is more pronounced than in simulations by Rogers & Pittard (2013), who modelled stellar winds and SNe feedback in a cloud with similar density but slightly higher mass ($3240 \, M_\odot$), containing three massive stars between 28 and 35 $M_\odot$. During the main-sequence phase, the mass flux does not exceed $5 \times 10^{-4} \, M_\odot \, \text{yr}^{-1}$, or a quarter of the peak flux in our model. Although their grid is twice the size of ours, meaning material must travel further to leave the grid, the mass flux is relatively uniform over the first 4 Myr, with fluxes only increasing after the evolved Wolf–Rayet and SN phases; fluxes higher than $2 \times 10^{-3} \, M_\odot \, \text{yr}^{-1}$ are only achieved after this point. This implies photoionization is a more efficient feedback mechanism for dispersing clouds than stellar winds, and is more comparable to SNe, lending support to Matzner (2002) who concluded that H II region expansion is the dominant source of feedback in GMCs. Rogers & Pittard (2013) found that dense gas was largely unaffected by feedback, with winds dispersing through low-density channels from the initial conditions. The resilience of dense gas was also found by Dale et al. (2012). This is borne out by our model as well.

The early hydrodynamical models of champagne flows in 1D and 2D by e.g. Tenorio-Tagle (1979), Bodenheimer, Tenorio-Tagle...
& Yorke (1979), and Yorke et al. (1989) – of an ionizing O star located next to the interior boundary of a molecular cloud – show the same characteristics as our 3D model presented here (see Fig. 5): the ionized cloud gas escapes as a champagne flow at 30 km s\(^{-1}\), while the diffuse ionized gas outside of the cloud expands at 10 km s\(^{-1}\). Such velocities are also observed in real H\,\textsc{ii} regions such as the Orion nebula (M42; O’Dell et al. 2017), DR 21 (Immer et al. 2014), and the Hourglass in M8 (Chakraborty & Anandarao 1997). The schematic in fig. 4 of O’Dell et al. (2009), interpreting observations of the Orion nebula, is remarkably similar to the champagne flow we see in Fig. 5, including the supersonic gas travelling away from the massive star, as well as the ionization front stalling at dense cores, from which photoevaporated material flows back towards the ionizing source.

The structures produced in our model are likely to be a result of the initial conditions. Walch et al. (2013) found similar structures in simulations where the initial conditions were filamentary and had high fractal dimension, whereas spherical H\,\textsc{ii} regions arose from conditions with lower fractal dimensions and more spherical geometries. Dale & Bonnell (2011) note that accretion flows towards deeply embedded massive stars limit the expansion of H\,\textsc{ii} regions, and therefore molecular clouds with such stellar distributions remain relatively undisturbed by feedback. On the other hand, when massive stars are located closer to the edge of molecular clouds, such as in the Orion nebula, their H\,\textsc{ii} regions may disperse gas effectively via champagne flows (Henney, Arthur & García-Díaz 2005). In our model, the massive star is located deep within the cloud, but it is still able to blow out a champagne flow through nearby low-density channels. Once the flow breaks out, the expanding H\,\textsc{ii} region is able to disrupt and disperse the rest of the cloud – even if the starting position was dense and more resistant.

One of the limitations of our model is that stars are placed in a particular location instead of being formed self-consistently via accretion; the star is positioned fully formed in a node between several filaments after the initial self-gravitating, turbulent evolution of the cloud. Therefore, we also do not model the evolution of an ultracompact H\,\textsc{ii} region alongside the growth of the star and the possible feedback effects this may incur. However, models by Peters et al. (2010) of massive protostars at the sub-parsec scale show that ultracompact H\,\textsc{ii} regions ‘flicker’, while the star is still accreting and only grow to substantial sizes after mass reservoirs are depleted. Therefore, as an approximation for scales of a few to 10 pc, a significant H\,\textsc{ii} region only blows out into the cloud once the star reaches its final mass, which is the stage we start with in our model. Furthermore, our calculation is informative for how the gas is displaced after this stage, and allows us to compare with, for example, models of stellar winds by Rogers & Pittard (2013) who place three massive stars in the centre of a turbulent medium. That said, for more comprehensive and self-consistent simulations, we plan to use sink particles to self-consistently grow star or cluster particles, using sub-grid models to compensate for the limited spatial resolution.

5.2 Temperature

The volume-average ionized gas temperature is approximately (9000 ± 1000) K over the course of the simulation. M42 (the Orion nebula) has a comparable electron temperature at (9200 ± 1600) K, with the fluctuation depending on the observational diagnostic (O’dell 2001).

Haworth et al. (2015) provided a temperature parametrization of the same thermal balance calculation as in our model, but for an H\,\textsc{ii} region expanding into an initially uniform-density medium. The ionized gas temperature is described by

\[ T_i = T_n + \eta \left[ 1.1 \times 10^4 - 3.8 \times 10^3 \left( \frac{z}{z_0} - 0.5 \right)^{0.839} - T_n \right] \]  

(26)

where \( T_n \) is a prescribed fully neutral gas temperature (e.g. 10 K or the dust temperature), \( \eta \) is the ionization fraction of hydrogen, and \( z \) is the metallicity relative to the Lexington benchmark metallicity \( z_0 \) (which we also use here; Ferland 1995; Ercolano et al. 2003; Haworth & Harries 2012). Haworth et al. used the same gas heating and cooling rates as our model, and so it accurately matches our volume-average temperature (9000 K; Fig. 8). (The two models used slightly different dust size distributions, so the gas-grain heat exchange term would be different; however, in the ionized gas, the rates of ionization heating and metal cooling would dominate over the gas-grain exchange rate.) This equation may be useful for those looking to use a simplified temperature scheme to account for the same thermal balance terms as our calculation, provided the ionization fraction is already known. However, it does not take into account the scatter in temperature which is about 10 per cent.

5.3 FUV interstellar radiation field

In Section 4.4, we showed how the FUV flux reaching a 0.88 M\(_\odot\) star increases rapidly as gas is removed from its surroundings. Photons at these wavelengths (912–2400 Å) cause photoelectric heating of dust and photodissociation of H\,\textsc{2} (Draine 1978; Osterbrock & Ferland 2006). Protoplanetary discs, or proplyds, around such stars in real star-forming regions are therefore stripped of material as the thermal pressure increases, resulting in a photoevaporative wind blowing from their outer layers (O’Dell, Wen & Hu 1993; Kim et al. 2016). This is not negligible, as external irradiation from other stars can be greater than internal irradiation by many orders of magnitude (Bruderer et al. 2012). Disc models typically include some external source of flux that remains constant in time (e.g. Haworth et al., 2016), but this is not representative of real clusters where stars move around and gas is displaced – a proplyd may see a radiation field that switches ‘on’ or ‘off’ depending on the intervening gas dynamics. This may in turn affect proplyd dispersal rates. In a future study, we intend to characterize the time-varying nature of the FUV flux around all 28 stars in our simulation.

6 SYNTHETIC OBSERVATIONS

Synthetic observations are produced using the temperatures, densities, dust properties, elemental abundances, and ionization fractions that were calculated and evolved during the RHD model – they were not modified with any post-processing. In this section, we analyse the model with both photoionization and radiation pressure.

6.1 Recombination and forbidden lines

We produce synthetic observations of the hydrogen recombination line H\,\alpha at 6563 Å, and collisionally excited metal forbidden lines of [S\,\textsc{ii}] at 6731 Å and of [O\,\textsc{iii}] at 5007 Å. Fig. 13 shows a three-colour composite at 0.2 Myr intervals corresponding to the column density snapshots in Fig. 4. Each line is scaled up to have the same minimum and maximum. As a representative example of the range of original surface brightnesses, at 0.6 Myr, the [S\,\textsc{ii}]
Figure 13. Three-colour images using synthetic observations of [S II]6731 Å (red), Hα (green), and [O III]5007 Å (blue), at 0.2 Myr intervals (corresponding to the column density maps in Fig. 4). The 34 M⊙ star is indicated by a point near the centre of each frame.

brightness varies between 0.03 and 9000 MJy str⁻¹; Hα lies between 0.3 and 3 × 10⁴ MJy str⁻¹; and [O III] has values between 0.1 and 10⁴ MJy str⁻¹. All three lines are strongest around clumps of high density, where cooling is more efficient, with the [S II] line showing the greatest difference in brightness between diffuse gas and dense gas. This causes the brown colour in the composite. From 0.4 Myr onwards, there are several clumps with bright-rimmed envelopes and tails pointing radially away from the massive star. These characteristics are shared with the proplyds, elephant trunks, and cometary knots seen in well-known H II regions spanning multiple size scales, such as the Orion nebula, NGC 7293 (the Helix, a planetary nebula; O’Dell 2000), the Carina nebula (Haikala et al. 2017), the Eagle nebula, and the Rosette cloud (Tremblin et al. 2013). We save a full characterization of masses and sizes for a future study, but values are of the order 1–30 M⊙ and 0.1–1 pc².

6.1.1 Hα line luminosity

In Fig. 14, we plot the intrinsic Hα luminosity \( L_{\text{int}} \) and the observed luminosity \( L_{\text{obs}} \) as a function of time, where

\[
L_{\text{int}} = 4\pi j_o \, dV
\]

and

\[
L_{\text{obs}} = 4\pi d^2 S_o
\]

where \( j_o \) is the emission coefficient, \( S_o \) is the Hα flux density, or surface brightness integrated over solid angle, and we have arbitrarily observed the model at a distance \( d = 400 \) pc (the distance to the Orion nebula) from three different viewing angles \( (\theta, \phi) = (0, 0), (90, 0), (90, 90) \) with the colatitude \( \theta \) and azimuthal angle \( \phi \) in degrees. Within the cloud itself, \( L_{\text{obs}} \) is reduced along the line of sight due to absorption and scattering by dust. The extinction

\[
A(\text{Hα}) \text{ in magnitudes is}
\]

\[
A(\text{Hα}) = -2.5 \log_{10} \left( \frac{L_{\text{obs}}}{L_{\text{int}}} \right)
\]

and this is also plotted in Fig. 14. The peak intrinsic luminosity, \( 8 \times 10^{37} \) erg s⁻¹, is reached at 0.2 Myr and this drops below \( 10^{35} \) erg s⁻¹ after a Myr. The extinction is a few magnitudes within the first 0.4 Myr, while the H II region is still embedded within the cloud, after which the extinction drops to zero as the region becomes optically thin.

Two schemes of scattering are considered. The first uses the peel-off method (see Yusef-Zadeh, Morris & White 1984): for photon packets which start off travelling away from the line of sight, their scattering events forcibly direct some light towards the observer (regardless of the new direction of the photon packet). This therefore adds scattered light from other lines of sight into the observed beam. Additionally, photon packets which do travel directly towards the observer may be scattered away from the observed beam. The second scattering scheme (labelled ‘non-scat’ in Fig. 14) only accounts for the latter effect – scattering from other lines of sight is not included.

The extinction between different viewing angles differs by 1–2 mag. Furthermore, the extinction with the full scattering treatment is lower by a magnitude compared to observations neglecting the peeled-off photons. This is because the emitting gas is surrounded by dense filaments of dust which has a scattering opacity peaking near 6563 Å (see Fig. 2), and this directs light towards the observer. Additionally, photon packets which do travel directly towards the observer may be scattered away from the observed beam. The second scattering scheme (labelled ‘non-scat’ in Fig. 14) only accounts for the latter effect – scattering from other lines of sight is not included.

The extinction between different viewing angles differs by 1–2 mag. Furthermore, the extinction with the full scattering treatment is lower by a magnitude compared to observations neglecting the peeled-off photons. This is because the emitting gas is surrounded by dense filaments of dust which has a scattering opacity peaking near 6563 Å (see Fig. 2), and this directs light towards the observer. Fluctuations in the extinction after 0.8 Myr arise from small differences in ionization; this has a more pronounced effect on the luminosity at late times as the luminosity is already dim, the ionized gas is diffuse, and recombination lines are sensitive to the square of the density.
Figure 14. Top – He luminosities as a function of time, with the intrinsic luminosity in solid black, and observed luminosities at viewing angles \((\theta, \phi) = (0, 0), (90, 0), (90, 90)°\) in blue, green, and red respectively. Observations without taking into account scattering from other lines of sight are in dotted lines, observations with scattering are dashed – see Section 6.1.1 for a full description. Bottom - extinctions for the aforementioned parameters.

6.2 Free–free radio observations and Lyman flux

It is possible to estimate the rate of Lyman continuum photons \((\lambda < 912\,\text{Å}; h\nu > 13.6\,\text{eV})\) being produced in a nebula using radio observations. Rubin (1968) derived an expression to calculate this assuming photoionization equilibrium, which implies

\[
\dot{N}_{\text{Ly}} \geq \int n_{i}(n(\text{H}\,\text{II}) + n(\text{He}\,\text{II})) \alpha(T)\,dV
\]  

(30)

where \(\alpha(T)\) is the recombination rate coefficient. Rubin uses the same \(\alpha\) for H and He, and to be consistent with the literature and to test the resulting diagnostic, we do the same in our analysis. This takes the form

\[
\alpha(T) = 4.10 \times 10^{-10} T^{-0.8}
\]

(31)

The optical depth for free–free radiation is

\[
r_{\text{ff}}^\nu = 8.235 \times 10^{-2} \left( \frac{T}{K} \right)^{-1.35} \left( \frac{\nu}{\text{GHz}} \right)^{2-\beta} \int \left( \frac{n_e n_i}{\text{cm}^{-6}} \right) \, dx \quad \text{pc}^{-1}
\]

(32)

Comparing with equation (30) leads to Rubin’s expression for the Lyman continuum production rate

\[
\dot{N}_{\text{Ly}} \geq 4.76 \times 10^{42} \left( \frac{\nu}{\text{GHz}} \right)^{0.45} \left( \frac{T_0}{K} \right)^{-0.45} \left( \frac{S_{\nu}}{\text{Jy}} \right) [\text{s}^{-1}]
\]

(34)

where \(S_{\nu}\) is the surface brightness integrated over the solid angle subtended by the object observed from a distance \(D\), \(T_0\) is the average temperature, \(\nu\) is the observation frequency, and \(\beta\) is a spectral index; this depends on the free–free Gaunt factor and is usually 0.1 in the literature (e.g. Mezger & Henderson 1967; Rubin 1968; Leffloch, Lazareff & Castets 1997; Kim et al. 2017). Equations (30) and (34) represent a lower limit in a density-bounded H\,\text{II} region.

We apply equation (34) to infer the Lyman continuum flux from synthetic observations of 20 cm free–free emission at 0.05 Myr intervals. We use a representative temperature of 9000 K as equation (34) is only weakly dependent on \(T_0\) and this is the average temperature in our simulation (see Fig. 8). The spectral index \(\beta\) is calculated to be 0.15. We plot the results in Fig. 15, along with the known production rate taken from the integrated stellar spectra. We also compare this with the equilibrium equation (30), which does not depend on any synthetic observations.

The radio measurement closely matches the result from equation (30) at all times, showing that the radio emission is accurately measured and tracks photoionization balance consistently. Before 0.2 Myr, both results underestimate the actual photon production rate by a factor of a few. Rubin’s method assumes that Lyman continuum photons only go into ionizing the gas, neglecting the number which are absorbed by dust grains and thermally re-emitted in the infrared, reducing the number available for the gas. Since the star at this stage is deeply embedded in a node between filaments, dust absorption is not negligible. Once the H\,\text{II} breaks out, however, the probe becomes more reliable, matching the actual emitted photon...
flux at 0.2 Myr. After this stage, the radio flux decreases along with the electron density, and by 0.8 Myr, the measured $N_{e_0}$ is an order of magnitude lower than the known production rate. By the end of the simulation, the discrepancy is $10^3$. Mass begins to leave the grid after around 0.4 Myr, so after this stage the ionizing photons may escape from the volume; if our cloud represents a core embedded inside a larger GMC, these photons could go on to excite gas beyond the model boundary. This would account for the increasing discrepancy at later times. Since the radio method is used by observers to get spectral classifications of O stars, this highlights the importance of knowing the full-size scale of an H II region — for a limited observational field of view, the Lyman flux can be underestimated by several orders of magnitude.

### 6.3 Dust emission

These synthetic observations were calculated using dust temperatures which are decoupled from gas, apart from the collisional heat exchange term in equation (12).

Dust surface brightness depends on the Planck function $B_\lambda(T_d)$ and in the optically thin limit is also proportional to the dust opacity. Taking the ratio of brightnesses at two wavelengths allows the dust temperature to be calculated if the opacity spectral index is known:

$$\frac{S_\lambda_1}{S_\lambda_2} = \left(\frac{\lambda_2}{\lambda_1}\right)^{3+\beta}\frac{\exp(hc/\lambda_2 k T_d) - 1}{\exp(hc/\lambda_1 k T_d) - 1}$$

where $S_\lambda_1$, $S_\lambda_2$ are the surface brightnesses at wavelengths $\lambda_1$, $\lambda_2$, and $\beta$ is the dust opacity spectral index. $\beta$ is normally taken to be 2 for regions such as the one modelled here (e.g. Reid & Wilson 2005; Sadavoy et al. 2012; Rumble et al. 2015; Figueira et al. 2017) and this is the value we use.

We apply equation (35) pixel-by-pixel to synthetic observations at 450 and 850 μm, corresponding to the Submillimetre Common User Bolometer Array on the James Clerk Maxwell Telescope. The resulting temperature map is presented in Figs 16 and 17, next to an intrinsic value ($T_d^{\text{int}}$) which is a dust-mass-weighted average along the line of sight calculated directly from the RHD model.

The observed temperature ($T_d^{\text{obs}}$) is accurately recovered in dense filaments. In warm diffuse gas, the dust temperature is underestimated by a factor of 2 and this gets worse as the H II region cavity grows and becomes less dense. This is where the most massive star is located and is where discrepancies are highest. Koepferl et al. (2017b) produced synthetic dust emission from SPH models of Dale et al. (2014) and concluded that errors in temperature were present in regions which had both low-density and high-temperature dispersion. We agree with this at early times when the H II is still mostly confined (Fig. 16). At later times when the H II region is extended and diffuse, inconsistencies are more correlated with low densities, with some high-temperature dispersion areas still being recovered fairly accurately (e.g. the filaments on the right of Fig. 17). They also note that their errors are found in places that are cooler than their surroundings and are greatest at the edge of the H II region. We do not in general find this to be the case, as the differences in our model are concentrated in the interior at both early and late stages; furthermore, some cool filaments entrenched in warmer areas still have accurate temperatures. Koepferl et al.’s method of SED fitting requires both the surface density and dust temperature to be free parameters, with overestimates in one leading to underestimates in the other. The method we employ here does not make any assumption on the density and therefore any discrepancies are discrepancies in temperature only. Using the ratio method only requires observations at two wavelengths, while the blackbody fitting method needs many wavelengths — temperatures in dense regions with low-temperature dispersion are accurately calculated using both methods, so in these areas observers may find the ratio method more useful due to its less stringent data requirements.

### 7 SUMMARY AND CONCLUSIONS

We have modelled a 1000 M⊙ cloud containing a 34 M⊙ massive star including photoionizing radiation and radiation pressure feedback. In summary:

(i) The cloud is dispersed within 1.6 Myr or 0.74 ($t_d$), with all mass leaving the (15.5 pc)$^3$ grid over this time.

(ii) Thermal pressure from photoionization is an efficient feedback mechanism, causing mass fluxes of the order of $10^{-3}$ M⊙ yr$^{-1}$ at the simulation volume boundary.

(iii) At most 40 per cent of the mass gets ionized (440 M⊙), while almost 90 per cent of the volume gets ionized. This arises from the densest filamentary structures resisting ionization, getting shovelled by the expanding ionization front into globules and pillars, which remain neutral and shield downwind material from the stellar radiation field.

(iv) Radiation pressure plays a negligible role, causing only a slight delay in the breakout of the ionization. It is expected to be more significant at higher number densities, $n > 100$ cm$^{-3}$ (Draine 2011).

(v) We use a detailed RT scheme in our models, calculating photoionization and thermal balance before each hydrodynamics time-step. Ionization fractions are calculated for multiple atomic species, and temperatures are calculated for gas and dust separately. These are then used to create self-consistent synthetic observations.

(vi) We also calculate the FUV interstellar radiation field, $G_0$, throughout the simulation volume, including around sink particles which can suddenly be exposed to the flux from the massive star as gas is dispersed.

Our synthetic observations include line and continuum emission. We have tested the use of radio free-free emission in probing the production rate of Lyman continuum photons. The ‘observed’ rate is almost always underestimated — by a factor of a few at early times (before 0.4 Myr), and by up to three orders of magnitude at late times (once significant amounts of mass have left the simulation volume, and thus photons would excite this gas beyond the boundary). We emphasize that radio measurements serve as a lower boundary on the production rate, especially when only part of the H II region is observed and if ionizing photon escape fractions are not also measured.

We also investigated the use of brightness ratios of synthetic dust continuum at two wavelengths (450 and 850 μm) to probe the dust temperature. This accurately recovers the actual temperature in regions of high-density and low-temperature dispersion. However, in low densities or high-temperature dispersions, the ‘observed’ temperature is underestimated by a factor of 2 or more, getting worse at late stages in the very diffuse H II region ($n < 3$ cm$^{-3}$). At high densities, the ratio method is as accurate as SED fitting which requires more than two wavelengths and has surface density as an additional free parameter that can cause further discrepancies in temperature (Koepferl et al. 2017b).

The cloud we have modelled here represents a cloud core as opposed to a GMC in its own right, therefore it would be embedded inside a larger mass object, which itself would be inside an even larger mass one. In future models we plan to go up the hierarchy to full GMC-scale regions of 10$^5$ M⊙ and 100 pc. Currently, we
place stars as sink particles in dense regions, but the particles do not accrete material (though they do move around with gas and N-body interactions). For the higher mass cloud models, we will implement a sub-grid model for sink particles such that they represent clusters of stars, and will enable accretion so that they grow self-consistently.

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