



Less wrong: a more realistic initial condition for simulations of turbulent molecular clouds

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ABSTRACT

Simulations of isolated giant molecular clouds (GMCs) are an important tool for studying the dynamics of star formation, but their turbulent initial conditions (ICs) are uncertain. Most simulations have either initialized a velocity field with a prescribed power spectrum on a smooth density field (failing to model the full structure of turbulence) or ‘stirred’ turbulence with periodic boundary conditions (which may not model real GMC boundary conditions). We develop and test a new GMC simulation setup (called TURBSPHERE) that combines advantages of both approaches: we continuously stir an isolated cloud to model the energy cascade from larger scales, and use a static potential to confine the gas. The resulting cloud and surrounding envelope achieve a quasi-equilibrium state with the desired hallmarks of supersonic ISM turbulence (e.g. density PDF and a $\sim k^{-2}$ velocity power spectrum), whose bulk properties can be tuned as desired. We use the final stirred state as initial conditions for star formation simulations with self-gravity, both with and without continued driving and protostellar jet feedback, respectively. We then disentangle the respective effects of the turbulent cascade, simulation geometry, external driving, and gravity/MHD boundary conditions on the resulting star formation. Without external driving, the new setup obtains results similar to previous simple spherical cloud setups, but external driving can suppress star formation considerably in the new setup. Periodic box simulations with the same dimensions and turbulence parameters form stars significantly slower, highlighting the importance of boundary conditions and the presence or absence of a global collapse mode in the results of star formation calculations.

Key words: MHD – turbulence – stars: formation – ISM: clouds – ISM: kinematics and dynamics – ISM: structure –.

1 INTRODUCTION

The initial conditions (ICs) for star formation (SF) are giant molecular clouds (GMCs), which are complex objects with a highly dynamic life cycle. Their evolution is regulated by supersonic magnetohydrodynamic (MHD) turbulence, with a substructured, filamentary morphology, and have energy continuously cascading from the turbulent eddies on the galactic scale height down to smaller scales (Mac Low & Klessen 2004; McKee & Ostriker 2007; Hopkins 2012; Padoan et al. 2016). GMCs form and disperse dynamically as part of the gas cycle in galaxies, exchanging mass and energy with the surrounding interstellar medium (ISM; Hopkins, Quataert & Murray 2012; Dobbs & Pringle 2013; Ibáñez-Mejía et al. 2016; Benincasa et al. 2020; Chevance et al. 2020; Guszejnov et al. 2020a). Star formation involves various non-linear, coupled physical processes acting on a large range of physical scales, so an important tool for understanding it is numerical simulations (e.g. Naab & Ostriker 2017;

Faucher-Giguère 2018; Teyssier & Commerçon 2019). However many different factors can potentially affect the outcome of SF, so the initial and boundary conditions for such simulations should be as realistic as possible to model the true complexity of GMC evolution.

An ideal solution would be to simulate the entire multiphase gas cycle within the galaxy with GMCs forming and dispersing dynamically (e.g. Dobbs 2008; Hopkins et al. 2012; Agertz et al. 2013; Hu et al. 2016), but this is not currently feasible at the level of detail needed to capture the formation of individual protostars and the emergence of the stellar initial mass function (IMF). IMF-resolving simulations are currently practical for either isolated GMCs or dense clumps, or a presumed small ‘patch’ within a larger GMC, with a total gas mass limited by computational capabilities. Lacking the greater galactic context of GMC formation, such simulations must assume some simplified initial and boundary conditions, generally informed by the phenomenology of supersonic ISM turbulence.

A popular approach (e.g. Bate, Bonnell & Bromm 2003) is to initialize a uniform-density sphere with a Gaussian random velocity field with a $\propto k^{-2}$ power spectrum (Dubinski, Narayan & Phillips 1995) to emulate turbulence (which we refer to as the SPHERE setup in this and related works – Guszejnov et al. 2020b; Grudić et al. 2021; Guszejnov et al. 2021). Such Gaussian ICs mimic real ISM

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turbulence, in so far as they model its 2-point velocity statistics initially (e.g. Larson 1981) by construction. Due to its simplicity, the SPHERE setup remains in use today as the most widely adopted initial condition for simulations of isolated star-forming gas clouds and cores, with various variations in the initial density profile employed (Girichidis et al. 2011; Rosen et al. 2016; Lee & Hennebelle 2018; Rosen et al. 2019), and typically with a uniform initial magnetic field when magnetic fields are included (e.g. Price & Bate 2007; Offner & Chaban 2017; Rosen & Krumholz 2020). Simulations of this type have been shown to reproduce certain aspects of GMCs found in self-consistent global galaxy simulations, in particular the density PDF once evolved from the initial uniform-density state (Rey-Raposo, Dobbs & Duarte-Cabral 2015).

However, fully developed turbulence is non-linear with high-order correlations between the density, velocity, and magnetic field. Such correlations are not captured by enforcing a certain initial velocity power spectrum.¹ Even if the power spectrum is tuned to reproduce observations, this does not imply a true turbulent cascade of energy from large scales down to the dissipation scale. Without ongoing energy injection, smaller eddies decay on their shorter turnover time-scale $\tau_{\text{edd}} \sim \lambda/v(\lambda) \sim \lambda^{0.5}$ (Gammie & Ostriker 1996; Mac Low 1999), changing the shape of the power spectrum over time. Lastly, the lack of material surrounding the cloud is not clearly realistic, as there would likely be some continuity between the material in the ‘cloud’ and the larger scale galactic gas flow that formed it. Therefore a more realistic IC is desirable.

Other simulations have modelled the initial turbulent state by ‘stirring’ turbulence in a periodic box for a certain length of time (e.g. Mac Low 1999), then switching on gravity to allow gravitational collapse and the formation of cores and/or stars from a self-consistent initial turbulent state, optionally with continued turbulent driving (Li et al. 2004). With continued driving, a true turbulent cascade with a steady power spectrum emerges (although it is still artificially truncated by numerical dissipation process at the resolution scale; Schmidt, Hillebrandt & Niemeyer 2004). Although it is still artificial to ‘switch on’ gravity instantaneously, such simulations (denoted BOX simulations in this work) are in a sense more realistic in both their initial state (determined by the emergent dynamics of turbulence), and in how they can model the ongoing injection of turbulent energy from larger scales not captured in the dynamic range of the simulation.

BOX-type simulations have sometimes arrived at different conclusions than equivalent SPHERE simulations for many important aspects of SF, including the rate of star formation, the shape of the IMF, the physical properties of cores and discs, stellar multiplicity, and the clustering of star formation (Offner, Klein & McKee 2008; Federrath & Klessen 2012; Krumholz, Klein & McKee 2012; Myers et al. 2014; Federrath 2015; Guszejnov et al. 2020b, 2021). Although such differences might be interpreted physically as the effect of more-realistic, self-consistent turbulence and/or driving, it is important to note that BOX models have differed from SPHERE models in multiple ways:

(i) Self-consistency of the initial turbulent state: SPHERE runs are generally run from approximate Gaussian initial conditions, while BOX runs begin with a self-consistent turbulent state.

(ii) Driving versus decay: SPHERE runs are typically run in a decaying configuration, whereas BOX runs often continue driving after gravity is switched on to model the cascade.

(iii) (Magneto-) hydrodynamic boundary conditions: SPHERE runs generally adopt vacuum boundary conditions or periodic boundary conditions with a box size much larger than the cloud (e.g. Price & Bate 2008), whereas BOX runs have periodic boundary conditions in a configuration where the cloud mass fills the whole box uniformly on average.

(iv) Gravitational boundary conditions: the solution to the Poisson equation in SPHERE simulations results in a confining central gravitational acceleration that is absent in a periodic BOX simulation. For a given set of turbulent conditions, BOX simulations can thus have an order-of-magnitude less gravitational energy (Federrath & Klessen 2012).

The boundary conditions of BOX setups may be problematic once feedback is introduced into the simulation: material that should physically be expelled from the cloud will instead run laps around the box (Nakamura & Li 2007), and even if feedback excites the box to a highly turbulent state the system cannot expand as it would according to the virial theorem. Mass-loss and disruption by feedback are a key phase of GMC evolution likely determining the end of star formation, but a GMC-sized periodic box setup cannot model such processes fully.

Moreover, a periodic solution to the Poisson equation freezes out the global gravitational mode, i.e. there is zero gravitational field in the absence of density perturbations, where physically the boundary conditions for any centrally concentrated gas structure would have a global confining potential. This global potential could potentially affect gas and stellar dynamics, the IMF (e.g. if there is a tendency for massive stars to sink to the global potential minimum and accrete gas; Bonnell et al. 2001), and star cluster structure and assembly (neighbouring subclusters are more likely to merge in a global confining potential; e.g. Bonnell, Bate & Vine 2003; Grudić et al. 2018). Practically by definition, GMCs are overdensities with respect to the surrounding ISM, so it could be more realistic to account for the confining potential that generically arises.

Thus we desire a simulation setup that has the best of both worlds: a setup that is localized in space with outflow or low-density boundary conditions like the SPHERE setup, but has the fully developed initial turbulence – and optional continued energy injection – of the BOX setup. In this work, we present a new type of simulation setup for GMCs, called TURBSPHERE, that has these properties. In Section 2 we describe our simulation code, our implementations of the SPHERE and BOX setups, and the new TURBSPHERE setup, and outline two suites of numerical simulations: one to investigate the behaviour of the TURBSPHERE turbulence stirring procedure, and the other including gas self-gravity to investigate the resulting star formation. In Section 3 we present the results of our simulation suites for different cloud parameters. In Section 4 we discuss the implications of our results regarding the roles of turbulence, continued driving, initial conditions, and boundary conditions in star formation simulations. In Section 5 we summarize our main findings, and in Appendix A we show the evolution of various equilibrium statistics for all of our TURBSPHERE stirring runs.

2 METHODS

2.1 Simulation code

We utilize the STar FORMation in Gaseous Environments (STAR-FORGE) numerical framework implemented within the GIZMO code

¹ Another notable, prior approach to Gaussian ICs is found in Klessen, Heitsch & Mac Low (2000), who used Lagrangian perturbation theory to generate an initial density structure consistent with a Gaussian random velocity field with a prescribed $\propto k^{-2}$ power spectrum, with periodic boundary conditions. This is nominally more realistic than assuming uniform density, but perturbation theory still cannot capture fully developed turbulence.

to perform our simulations. A full discussion and presentation of the STARFORGE framework, along with tests and algorithm details, is available in Grudić et al. (2021) (hereafter [Paper I](#)). Therefore, we only summarize the physics modules utilized below.

2.1.1 Core physics

We use the quasi-Lagrangian mesh-less finite-mass (MFM) method for magnetohydrodynamics (MHD) from Hopkins & Raives (2016). We assume ideal MHD with a constrained-gradient scheme that ensures $\nabla \cdot \mathbf{B} = 0$ to high precision (Hopkins 2016). For simplicity, all simulations in this work assume an isothermal equation of state $P = \rho c_s^2$, where $c_s = 0.2 \text{ km s}^{-1}$ is the typical isothermal sound speed in $\sim 10 \text{ K}$ molecular gas.

Gravity is solved with an improved version of the parallel Barnes–Hut tree solver from Springel (2005) combined with high-order integration of sink particle trajectories to accurately track multiple-star systems ([Paper I](#)). We optimize the gravity calculation for gas using the Grudić (2021) adaptive force updating scheme. Softening is fully adaptive for gas cells, while sink particles (representing individual (proto)stars) adopt a softening radius of 21 AU. Sinks follow the STARFORGE formation and accretion algorithm derived from that of Bate, Bonnell & Price (1995), requiring multiple checks to form a sink particle, which accretes other Lagrangian gas cells based on criteria accounting for magnetic, kinetic and gravitational energies and angular momentum.

Upon forming, sinks follow the protostellar evolution model described in Offner et al. (2009), treating the stars as collapsing polytropes with distinct phases. These phases (‘pre-collapse’, ‘no burning’, ‘core deuterium burning at fixed temperature’, ‘core deuterium burning at variable temperature’, ‘shell deuterium burning’, and ‘main sequence’) allow for dynamic evolution of the stellar properties, in particular the protostellar radius that sets the jet velocity.

2.1.2 Protostellar jets

Protostars eject a significant portion of their accreting mass in bipolar jets (Bally 2016; Pudritz & Ray 2019). These jets have been shown to suppress star formation globally and locally, significantly reducing the overall SF efficiency and the mass scale of individual stars (Hansen et al. 2012; Federrath et al. 2014; Rosen & Krumholz 2020; Guszejnov et al. 2021; Mathew & Federrath 2021). As these effects can be so important, we account for protostellar jet feedback in a subset of our star formation simulations. We use the parametrization of Cunningham et al. (2011), wherein a fraction f_w of the material accreting near the protostar is launched in a bipolar collimated jets along the sink angular momentum axis, with velocity

$$v_{\text{jet}} = f_K \sqrt{\frac{GM_\star}{R_\star}}, \quad (1)$$

where f_K represents the fraction of Keplerian velocity at the protostellar radius R_\star . We assume $f_w = f_K = 0.3$, setting the momentum loading parameter $f_w f_K$ in the range inferred from observations and in line with other works (Cunningham et al. 2011; Offner & Chaban 2017). Jet feedback is implemented numerically by ‘spawning’ new Lagrangian gas cells near the sink particle, where cells are launched in a narrow cone about the axis of angular momentum of the sink particle. Further in depth discussion on the protostellar jets implementation can be found in the STARFORGE methods paper ([Paper I](#)).

2.1.3 Turbulence driving

For simulations with external driving, we use the turbulence driving scheme introduced in Schmidt et al. (2004). We stir gas throughout the box with a body force considering of Fourier modes corresponding to wavelengths ranging from 0.25 to 1 times the size scale of the simulation (cloud radius for isolated clouds, box size for periodic BOX ICs), with a peak amplitude at $\lambda = 0.5 R_0$ by default. These modes are generated in Fourier space as an Ornstein–Uhlenbeck process with a coherence time-scale of roughly 1 global free-fall time ($\tau = 3 \text{ Myr}$). The compressive part of the acceleration is projected out via a Helmholtz decomposition in Fourier space so that the driving is purely solenoidal (Federrath, Klessen & Schmidt 2008), modelling the effect of large-scale supernova driving on GMC scales (Padoan et al. 2016).

2.2 Initial conditions and setups

In this work, we perform simulations with three different choices of initial and boundary conditions, comparing SPHERE and BOX setups (which are commonly employed in star formation studies) to our new TURBSPHERE setup, which we detail below. Fig. 1 shows density projections of each simulation set-up at initialization, which we describe in following sections. In all simulations we adopt a mass resolution $\Delta m = 10^{-3} M_\odot$, resolving an $M_0 = 2 \times 10^4 M_\odot$ cloud with 2×10^7 equal-mass gas cells. While our simulations have adaptive cells (and thus do not have fixed sizes), the roughly equivalent maximum spatial resolution of our setup is on the order of 30 AU ([Paper I](#)).

2.2.1 SPHERE

Our SPHERE IC consists of a spherical cloud of uniform density with mass $M_0 = 2 \times 10^4 M_\odot$ and radius $R_0 = 10 \text{ pc}$, positioned in the centre of a periodic box of side length $10 R_0$. The cloud is enveloped by a diffuse medium with a 1/1000 density contrast. The initial magnetic field is uniform, with its strength B_0 set by the chosen dimension-less mass-to-flux ratio μ_0 , the ratio of magnetic flux through the cloud to the critical flux determined in (Mouschovias & Spitzer 1976):

$$\mu_0 = 0.4 \frac{G^{1/2} M_0}{\pi R_0^2 B_0} = 0.42 \sqrt{\frac{|E_{\text{grav}}|}{E_{\text{mag}}}}, \quad (2)$$

where E_{grav} and E_{mag} are the gravitational and magnetic energy of the gas within the cloud, and $\mu_0 = 4.2$ is our fiducial value (equivalent to $2 \mu \text{G}$), consistent with the observed magnetic field strengths of dense molecular gas (Crutcher et al. 2010). Finally, to mimic the velocity correlations of turbulence, an initial velocity field is initialized (in agreement with Burkert & Bodenheimer 2000), where a Gaussian random field (with a power spectrum of $P_v(k) \propto k^{-2}$) is generated on a Cartesian grid and interpolated on to the cell positions. The velocity is normalized to achieve an initial virial parameter $\alpha_{\text{turb}} = 5 R_0 \mathcal{M}^2 c_s^2 / (3 G M_0) = 2$. However, $\alpha_{\text{turb}} \sim 1$ during most of the star formation due to the decay of turbulence.

2.2.2 BOX

In the BOX setup, gas cells are initially deposited on a uniform-density glass configuration in a periodic box with zero velocity. Similar to other runs in this suite, BOX runs have a mass $M_0 = 2 \times 10^4 M_\odot$ and have a size set to have the same volume-density ($L_{\text{box}} = 16.1 \text{ pc}$) as SPHERE runs. During initialization, gravity is disabled, and the

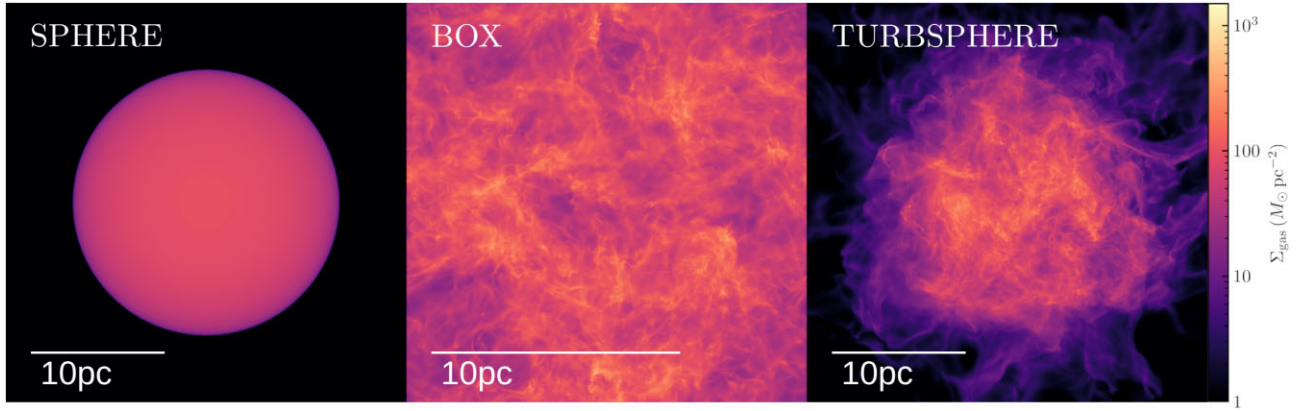


Figure 1. Surface density maps of our initial condition setups. SPHERE (left-hand panel) is a uniform sphere with a pre-generated, pseudo-turbulent $\propto k^{-2}$ velocity power spectrum. BOX (middle) shows the periodic-box setup, where fully realized, self-consistent turbulence is simulated before self-gravity is enabled. TURBSPHERE (right-hand panel) features a roughly spherical cloud with self-consistent turbulence resulting from stirring the gas while confining it with an ad-hoc potential.

periodic box is continually stirred by an injection of energy with a pre-determined power spectrum as described in Section 2.1.3, tuned to achieve an RMS Mach number $\mathcal{M} = \sqrt{\langle v^2/c_s^2 \rangle} \sim 10$. This is similar to the Mach number achieved in our fiducial TURBSPHERE setup and the Mach number in our SPHERE setup at the time that most stars form. The BOX IC initialization transforms an initial uniform magnetic field, via self-consistent evolution, to a magnetic field that mimics MHD turbulence.

After five freefall times, gravity is enabled. Thus, simulations using the BOX IC start with self-consistent MHD turbulence, wherein the resulting density distribution is nearly lognormal (Vazquez-Semadeni 1994; Padoan, Nordlund & Jones 1997) and the density, velocity, and magnetic fields have non-linear correlations.

2.2.3 TURBSPHERE

TURBSPHERE initial conditions are generated by invoking a SPHERE initial condition setup, introducing an analytic potential, and enabling driving to initialize a self-consistent turbulent state. The initial magnetic field is uniform with $\mu_0 = 4.2$ by default.

The turbulence initialization procedure uses an analytic potential – as opposed to the full self-gravity solution – because we wish to initialize the turbulent state self-consistently (as is also done in our BOX setup) without allowing stars to form at first. We have also experimented with two alternate approaches for modelling gravity while preventing SF: using a large gravitational softening length ($\epsilon \gtrsim 1$ pc), and making the gravity solution spherically symmetric while evolving the MHD equations in 3D. Using a large softening still allowed the spontaneous formation of bound structures and runaway collapse, while the spherically symmetric solver led to large, unstable swings in the cloud properties and either runaway collapse or cloud dispersal. We seek an initial condition close to some notion of equilibrium, so we adopt a static potential.

Inside the nominal cloud radius, R_0 , the analytic gravitational field is that of a homogeneous sphere matching the nominal cloud properties ($R_0 = 10$ pc, and mass $M_{\text{well}} = 2 \times 10^4 M_\odot$, unless we explicitly vary M_{well} to test the sensitivity of this parameter). Within $r < R_0$, the radial gravitational field is

$$g_r = -\frac{GM_{\text{well}}r}{R_0^3}. \quad (3)$$

Outside R_0 , we add an additional component modelling the field of mass halo with a $\rho \propto r^{-3}$ density profile, matched in density to the uniform sphere at $R = R_0$. Hence for $r > R_0$, the field is given by

$$g_r = -\frac{GM_{\text{well}} \left(1 + 3 \ln \left(\frac{r}{R_0} \right) \right)}{r^2}. \quad (4)$$

The inclusion of the additional halo component gives the potential infinite escape speed (to prevent mass-loss) without imposing a sharp confining barrier, because $\rho \propto r^{-3}$ is the steepest power-law mass model with infinite escape speed. During the turbulence initialization, gas cells are only subject to the analytic potential well and do not self-interact via gravity.

A driving force is applied as described in Section 2.1.3 as the simulation progresses, similar to the BOX case. The driving scheme is identical to that of BOX runs, and we tune the driving strength to achieve a nominal $\mathcal{M} \sim 10$, which is chosen to match SPHERE runs at the time that SF begins in those runs. We run the initialization until the gas reaches a quasi-equilibrium state, as we will demonstrate in Section 3.

2.3 Simulation suites

We run two suites of simulations in order to map the parameter space of possible TURBSPHERE initial conditions and determine the effects on subsequent star formation.

The first suite aims to investigate how the TURBSPHERE initialization setup behaves qualitatively, and how varying its free parameters affects the resulting dynamical equilibrium. We perform eight TURBSPHERE initialization runs, varying the scaling factor of the forcing strength \tilde{f} , the scaling factor of the wavelength range of our driving scheme $\tilde{\lambda}$, the strength of the analytic gravity well, and the mass-to-flux ratio μ_0 . All runs are listed in Table 1 and discussed in full in Section 3.1.

We then run eight different star formation simulations to determine how simulations from the new TURBSPHERE setup behave, and to identify the effects of protostellar jets, the turbulent cascade, simulation geometry, external driving, and gravity/MHD boundary conditions on the subsequent star formation statistics. We perform two SPHERE runs with and without protostellar jets, two driven BOX runs with and without jets, and four TURBSPHERE runs, switching driving and protostellar jets in each possible combination. We list

Table 1. List of TURBSPHERE turbulence initialization run parameters and results. Columns give (1) the normalization of the turbulent forcing relative to the fiducial run, (2) the range of wavelengths of driving modes in units of the initial cloud radius R_0 , (3) the mass sourcing the analytic gravity well in units of the nominal cloud mass M_0 , (4) the initial cloud mass-to-flux ratio, and the final equilibrium values of the cloud 3D RMS Mach number \mathcal{M}_{RMS} , half-mass radius R_{50} , RMS distance from centre of mass R_{RMS} , ratio of magnetic to kinetic energy in the cloud, and cloud virial parameter accounting for the cloud self-gravity α_{turb} (Columns 5–9 respectively). Equilibrium values are measured over the last 10 crossing times, and we also quote their $\pm\sigma$ variance over this interval.

Forcing strength	Forcing wavelength range (R_0)	$M_{\text{well}} (M_0)$	μ_0	\mathcal{M}_{RMS}	R_{50}/R_0	R_{RMS}/R_0	$E_{\text{mag}}/E_{\text{kin}}$	$\alpha_{\text{turb}} = 2E_{\text{kin}}/ E_{\text{self-gravity}} $
1	[1, 0.25]	1	4.2	10.9 ± 0.3	0.77 ± 0.05	0.88 ± 0.04	0.31 ± 0.03	1.0 ± 0.1
0.7	[1, 0.25]	1	4.2	9.5 ± 0.5	0.63 ± 0.06	0.72 ± 0.05	0.27 ± 0.03	0.6 ± 0.1
1.4	[1, 0.25]	1	4.2	12.1 ± 0.4	0.90 ± 0.04	1.09 ± 0.06	0.36 ± 0.03	1.7 ± 0.1
1	[1, 0.25]	0.5	4.2	9.8 ± 0.2	1.06 ± 0.05	1.34 ± 0.03	0.38 ± 0.03	1.7 ± 0.1
1	[1, 0.25]	2	4.2	11.5 ± 0.7	0.51 ± 0.04	0.58 ± 0.02	0.25 ± 0.04	0.62 ± 0.09
1	[2, 0.50]	1	4.2	14.3 ± 0.4	1.10 ± 0.09	1.64 ± 0.04	0.22 ± 0.02	3.6 ± 0.3
1	[1, 0.25]	1	1.3	10.5 ± 0.4	0.73 ± 0.04	0.84 ± 0.02	0.66 ± 0.06	0.89 ± 0.08
1	[1, 0.25]	1	0.42	10.4 ± 0.2	0.89 ± 0.02	1.05 ± 0.01	3.2 ± 0.1	1.07 ± 0.03

Table 2. List of star formation runs with their initial conditions and parameters. Columns give (1) the star formation simulation name, (2) the initial condition method used, (3) if driving is enabled, (4) if protostellar jets are enabled, (5) the initial cloud mass (M_0), (6) the initial mass-to-flux ratio (μ_0), (7) the 3D RMS Mach number (\mathcal{M}_{RMS}), (8) the size of the initial condition (R_{50} for TURBSPHERE & SPHERE runs, L_{box} for BOX runs), (9) the initial virial parameter, (10) and the final star formation efficiency, $\text{SFE}_{\text{final}}$.

Star formation simulations	Method	Driving	Jets	$M_0 (M_\odot)$	μ_0	\mathcal{M}_{RMS}	Size (pc)	α_{turb}	$\text{SFE}_{\text{final}}$
TURBSPHERE	TURBSPHERE	No	No	2×10^4	1.1	11.2	7.7	1.1	0.22
TURBSPHERE+Jets	TURBSPHERE	No	Yes	2×10^4	1.1	11.2	7.7	1.1	0.11
TURBSPHERE+Driving	TURBSPHERE	Yes	No	2×10^4	1.1	11.2	7.7	1.1	0.15
TURBSPHERE+Driving+Jets	TURBSPHERE	Yes	Yes	2×10^4	1.1	11.2	7.7	1.1	0.06
BOX+Driving	BOX	Yes	No	2×10^4	4.2	11.4	10	22	0.07
BOX+Driving+Jets	BOX	Yes	Yes	2×10^4	4.2	11.4	10	22	0.01
SPHERE	SPHERE	No	No	2×10^4	4.2	16.1	7.9	2.0	0.29
SPHERE+Jets	SPHERE	No	Yes	2×10^4	4.2	16.1	7.9	2.0	0.11

the initial conditions and parameters of the star formation runs in Table 2 and discuss their results in Section 3.2.

3 RESULTS

3.1 TURBSPHERE stirring runs

From the initial SPHERE setup state, the TURBSPHERE initialization runs continually inject energy into the system via the driving scheme discussed in Section 2.1.3 through their entire duration, until quasi-equilibrium is established. To ensure that this occurs, we ran TURBSPHERE initialization runs for $\sim 30t_c$, where

$$t_c = \frac{R_0}{c_s \mathcal{M}} \quad (5)$$

is the turbulent crossing time. While we run the simulations for several tens of t_c , an apparent quasi-equilibrium is established in a similar number of crossing times to BOX simulations ($2 \lesssim t/t_c \lesssim 3$, as found in previous works, e.g. Li et al. 2004).

The evolution of the cloud over many crossing times can be seen in Fig. 2, where the fiducial TURBSPHERE stirring run retains a relatively spherical shape, while a halo of diffuse material (made primarily of the original diffuse envelope from the SPHERE initial conditions and ejected matter) persists around the cloud. The cloud appears to be in quasi-equilibrium, with its macroscopic properties remaining steady over time.

Analysis of the system’s macroscopic properties, shown in Fig. 3, reveals that the size of the cloud is relatively stable throughout the entire evolution after ~ 2 crossing times, as evident by the stability of the ratio of the mass-weighted RMS and median radii R_{RMS} and R_{50} to the initial cloud radius R_0 . We also plot the virial parameter α_{turb} (neglecting the imposed potential well and accounting for the

binding energy the gas would have if self-gravity were enabled), the 3D RMS Mach number, and the ratio of magnetic to kinetic energy.

Despite the continual injection of energy into the cloud, all quantities of interest remain relatively stable after equilibrium is established due to the ongoing dissipation of turbulence in shocks (Mac Low 1999) and the confinement of the central cloud by the ad-hoc potential.

3.1.1 Turbulence statistics

To investigate the properties of the turbulence that develops in TURBSPHERE runs, we use our BOX initialization run as a reference, as the properties of MHD turbulence in this type of setup have been documented extensively (Mac Low 1999; Li et al. 2004; Federrath et al. 2008). The stirred TURBSPHERE cloud contains clumps and filaments with a morphology similar to those found in BOX clouds, as visually apparent in Fig. 1. We compare the mass-weighted density probability distribution functions (PDFs) of the fiducial stirring runs of both BOX and TURBSPHERE setups in Fig. 4.

Most of the mass in both PDFs is found in an approximately log-normal component, as expected generically in isothermal supersonic turbulence (Vazquez-Semadeni 1994; Nordlund & Padoan 1999). The additional low-density peak found in the TURBSPHERE density PDF is the diffuse, box-filling medium surrounding the cloud, and between the cloud and diffuse medium densities there is material in a halo surrounding the cloud. The TURBSPHERE and BOX density PDFs agree best at the highest densities, to the point of being practically indistinguishable in Fig. 4. This suggests that the two setups have overall similar density structure in the central regions of the cloud.

In Fig. 5, we show the velocity power spectrum $P_v(k)$ in the reference TURBSPHERE and BOX runs when turbulence is fully developed. Note that we expect a ‘bottleneck’ effect in the power

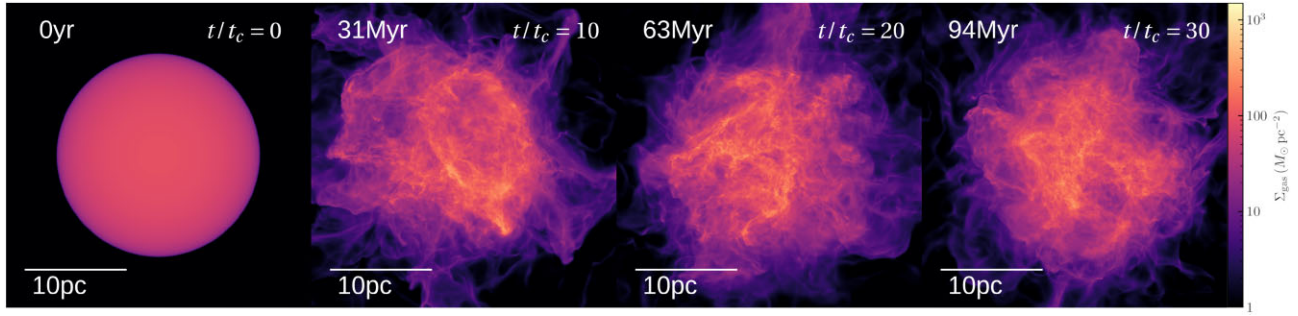


Figure 2. The evolution of our fiducial TURBSPHERE run, taken at intervals every 10 crossing times. After quasi-equilibrium is established, the cloud continues to evolve and sustain realistic turbulence through large-scale energy injection via driving.

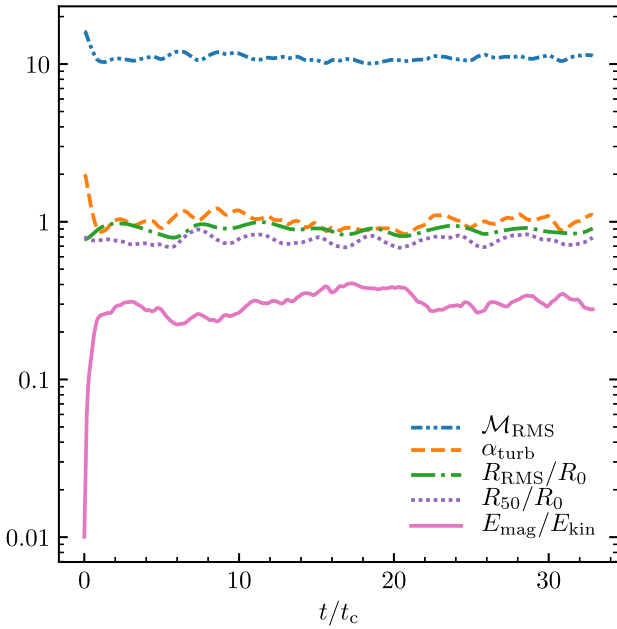


Figure 3. Evolution of our fiducial TURBSPHERE initialization run, plotting the 3D RMS Mach number \mathcal{M}_{RMS} , half-mass radius R_{50} , RMS distance from centre of mass R_{RMS} , ratio of magnetic to turbulent energy $E_{\text{mag}}/E_{\text{turb}}$, and virial parameter α_{turb} as a function of time in units of the crossing time $t_{\text{cross}} = R_0/(c_s \mathcal{M}_{\text{RMS}})$. All of these quantities approach equilibrium after 2–3 crossing times.

spectrum on scales comparable to the mean cell spacing $\Delta x \sim 0.1$ pc where the numerical method imposes an artificial steepening of the power spectrum (Schmidt et al. 2004; Padoan et al. 2007), and based on previous tests of supersonic turbulence in GIZMO at comparable resolution (Hopkins 2015) we expect this to occur at a physical angular wavenumber of $k \sim 100/L_{\text{box}}$ in our simulations. Such steepening is evident in both the TURBSPHERE and BOX power spectra for $k \gtrsim 10 \text{ pc}^{-1}$, so the physical results are at wavenumbers lower than this. In this range, the BOX run exhibits a power-law slope somewhat shallower than $\propto k^{-2}$ expected for unmagnetized supersonic turbulence, and closer to the $\propto k^{-5/3}$ law found in other similar supersonic MHD turbulence simulations (Li et al. 2004). The TURBSPHERE setup has a somewhat steeper spectral index of ~ -1.8 . This indicates that the choice of boundary conditions and confining potential can generally affect the velocity power spectrum measured in MHD turbulence simulations, even when controlling for \mathcal{M} , μ_0 , and the driving scale. Both setups’ power spectral indices lie within

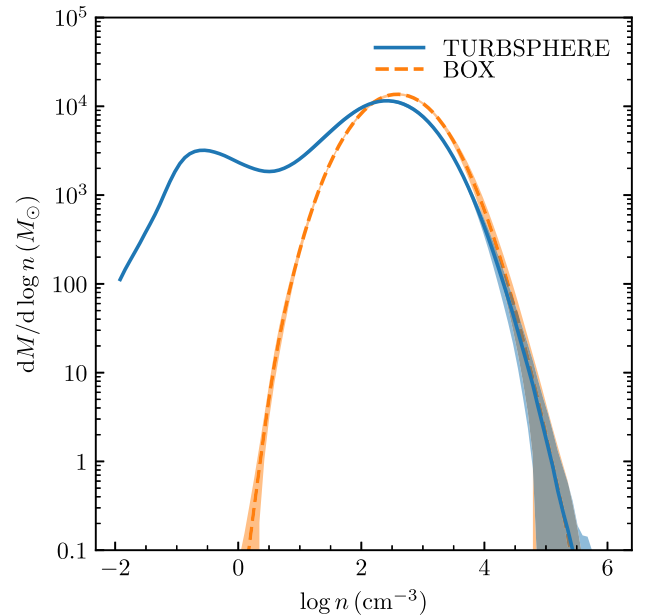


Figure 4. Mass-weighted density PDF in the fiducial stirring run, averaged over the final three crossing times with shaded interval indicating the standard deviation during this time due to intermittency. For comparison we also plot the density PDF from an $\mathcal{M} \sim 10$ driven-turbulence run in a periodic box with roughly equal volume-averaged density, averaged over the final three crossing times.

the range of values that have been inferred in real GMCs (Brunt & Heyer 2002; Padoan et al. 2003, 2006).

We also plot the magnetic field power spectrum $P_{\text{mag}}(k)$ in Fig. 6. The power spectra of the two setups are again qualitatively similar, with TURBSPHERE runs having a similar-yet-shallower slope for all values of k , which is opposite of what is seen in the velocity power spectra. In all, the basic density, velocity, and magnetic field statistics of the TURBSPHERE setup are fairly similar – but not identical – to those found in the BOX setup. Therefore, the turbulent properties of our TURBSPHERE setup demonstrates that, similar to the BOX method, this new method accurately models self-consistent supersonic turbulence in GMCs for star formation simulations.

3.1.2 Sensitivity to parameters

We note the equilibrium values of various statistics (\mathcal{M}_{RMS} , R_{50} , R_{RMS} , $E_{\text{mag}}/E_{\text{kin}}$, and α_{turb}) depend on parameters such as the forcing strength scaling factor \tilde{f} , driving wavelength range scaling factor

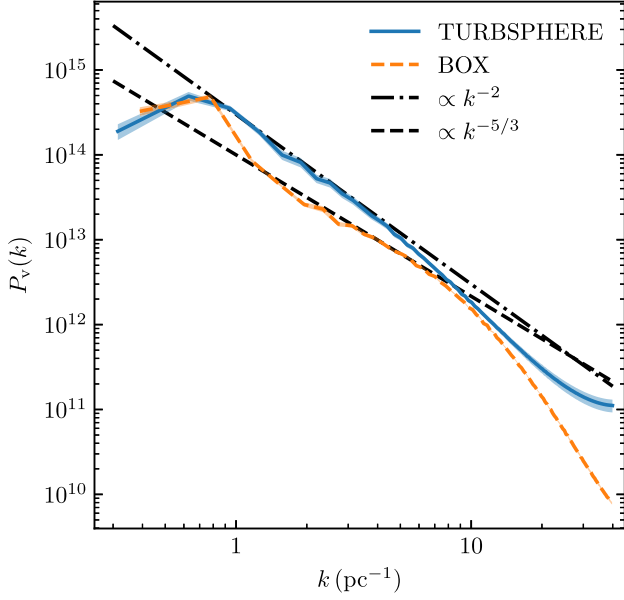


Figure 5. The velocity power spectrum of the the fiducial TURBSPHERE run, expressed as a power spectrum of the velocity field for given physical wavenumbers $k(\text{pc}^{-1})$, compared to that of the equivalent BOX run and a $\propto k^{-2}$ power law. We plot the average statistics of the last five crossing times and their standard deviations (shaded regions).

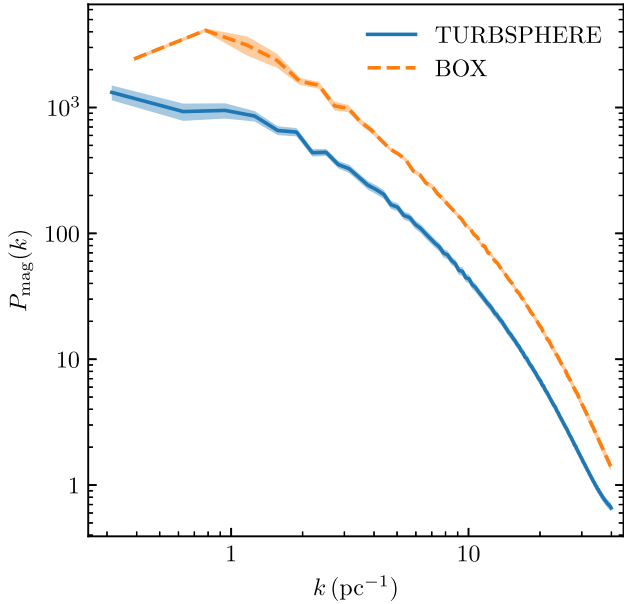


Figure 6. The magnetic field power spectrum of the fiducial TURBSPHERE run, expressed as power spectrum of the magnetic field for given physical wavenumbers $k(\text{pc}^{-1})$, compared to that of the equivalent BOX run. We plot the average statistics of the last five crossing times and their standard deviations (shaded regions).

$\tilde{\lambda}$, analytic gravity well scaling factor M_{well}/M_0 , and initial mass-to-flux ratio μ_0 in the TURBSPHERE stirring runs, listed in Table 1. To visually assess the impact of our choice of initial parameters, we plot the various equilibrium statistics of all stirring runs listed in Table 1 and place them in Appendix A. Unweighted least-squares logarithmic fits yield the following power-law relations that approximate all of our results to within $\sim 0.1\text{dex}$:

$$\frac{R_{50}}{R_0} \approx 0.82 \tilde{f}^{0.5} \tilde{\lambda}^{0.5} \left(\frac{M_{\text{well}}}{M_0} \right)^{-0.5}, \quad (6)$$

$$\frac{R_{\text{RMS}}}{R_0} \approx 0.97 \tilde{f}^{0.6} \tilde{\lambda}^{0.8} \left(\frac{M_{\text{well}}}{M_0} \right)^{-0.6}, \quad (7)$$

$$\frac{E_{\text{mag}}}{E_{\text{turb}}} \approx 1.29 \tilde{f}^{0.4} \tilde{\lambda}^{-1} \mu_0^{-0.8} \left(\frac{M_{\text{well}}}{M_0} \right)^{-0.3}, \quad (8)$$

$$\alpha_{\text{turb}} \approx \tilde{f}^{1.5} \tilde{\lambda}^{1.6} \left(\frac{M_{\text{well}}}{M_0} \right)^{-0.7}, \quad (9)$$

$$\mathcal{M} \approx 11 \tilde{f}^{0.4} \tilde{\lambda}^{0.4}. \quad (10)$$

The above approximations neglect all powers shallower than 0.1, and at this level of significance the only quantity that is sensitive to μ_0 is the magnetic energy itself, while the Mach number \mathcal{M} is insensitive to μ_0 and the properties of the confining potential. Using these formulae, an initial condition with any desired bulk equilibrium properties may be generated (within their respective domains of validity).

3.2 Star formation runs

We now examine the results of the BOX, SPHERE, and TURBSPHERE simulations run with various combinations of physics (see Table 2). In Fig. 7 we compare the star formation histories of each run with their star forming BOX and SPHERE counterparts. For each simulation, we plot the star formation efficiency ($=\sum M_*/M_0$), number of stars formed N_* , and the 3D Mach number \mathcal{M} , as a function of time.

We find general trends in star formation histories that differ between BOX, SPHERE, and TURBSPHERE simulations. BOX runs consistently show lower star formation efficiencies at fixed time, while SPHERE and TURBSPHERE have generally more-similar SFE. Furthermore, the slope of the SFE time evolution in BOX runs is shallower (SFE $\propto (t - t_0)^2$, e.g. Murray & Chang 2015; Lee, Chang & Murray 2015; Murray et al. 2017), while SPHERE and TURBSPHERE runs have steeper SFE slopes (similar to the $\propto (t - t_0)^3$ law found in previous SPHERE runs, e.g. Guszejnov et al. 2020b, 2021). A similar trend is also reflected in the number of stars N_* , where BOX simulations exhibit a shallower SF history than the others. We further discuss the physical explanation of the faster star formation rate (SFR) in Section 4.1.

Through the duration of the BOX simulations, we find a general stability in the Mach number. TURBSPHERE+Driving runs follow this general-stability trend, with divergence dependent on the presence of protostellar jets. In contrast, turbulence in the SPHERE runs decays quickly until feedback begins to inject significant kinetic energy into the cloud, which helps sustain turbulence (Rosen & Krumholz 2020; Appel et al. 2021). TURBSPHERE runs without driving also follow this trend, with Mach numbers initially declining until feedback replenishes turbulence. The effects of driving and the Mach number are further discussed in Section 4.2.2.

We plot the stellar mass functions (specifically, the quantity $dN/d\log M_*$) measured at the end of each simulation in Fig. 8. As found previously in BOX and SPHERE simulations (Guszejnov et al. 2020b), TURBSPHERE simulations without any feedback result in a mass distribution that is much too shallow top-heavy, with most stellar mass contained in stars more massive than $20M_\odot$. We do note some diversity in the shapes of the IMFs among the runs without feedback: the TURBSPHERE run peaks at a factor of 3–4 lower than the TURBSPHERE+Driving run, and the BOX+Driving and SPHERE run mass functions peak at even greater masses.

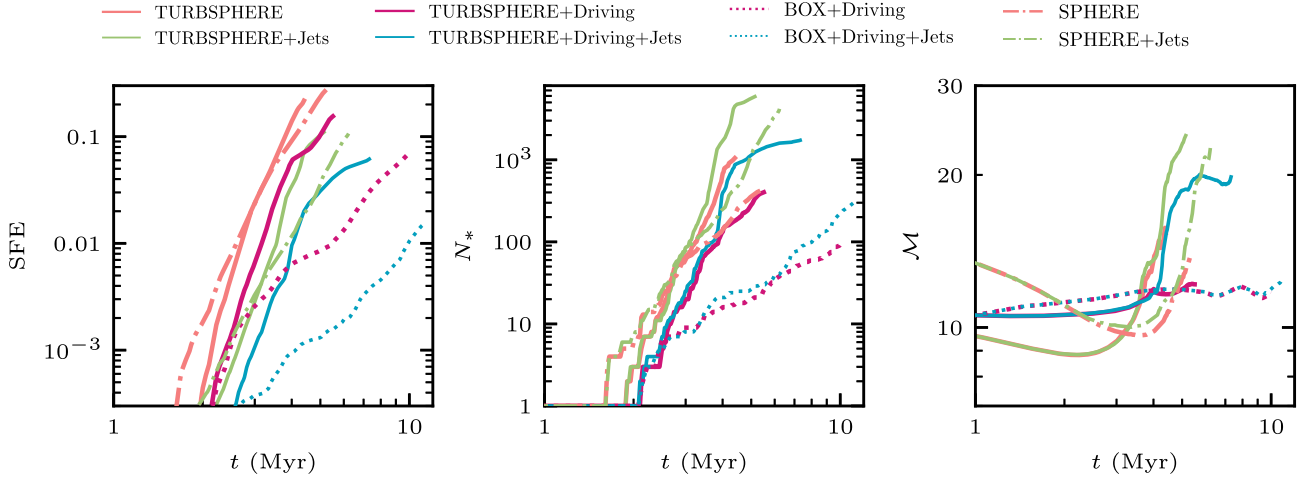


Figure 7. Statistics of the various TURBSPHERE, BOX, and SPHERE star formation runs with combinations of Jets and Driving. The initial condition setup of the TURBSPHERE runs are discussed in Section 3.2, with initial condition parameters from the runs in Table 2. Panels show (1) the star formation efficiency and (2) the number of stars formed, (3) and the 3D Mach number \mathcal{M} as a function of time (Myr).

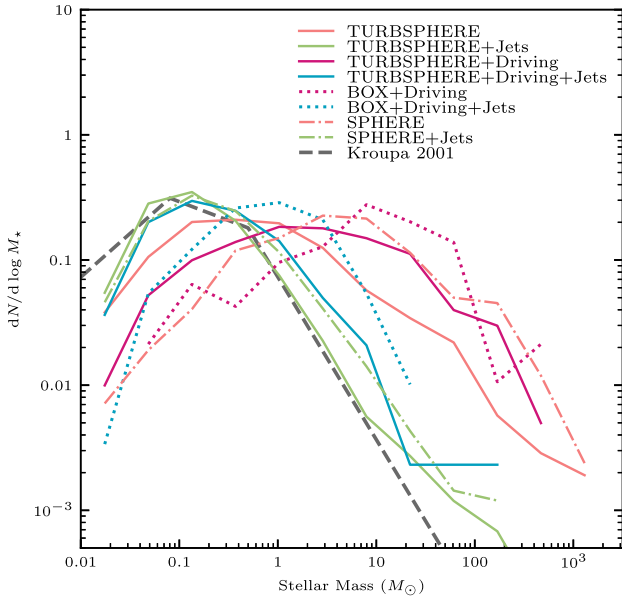


Figure 8. The stellar initial mass function (IMF) from the final snapshot of each star formation simulation with different physics and setups (see Table 2). We normalize each IMF and label the initial condition methods by line style and physics modules by colours. For comparison we plot the Kroupa (2001) IMF.

In agreement with prior work, we find protostellar jets reduce both the mass-scale of stars and the overall star formation efficiency (Hansen et al. 2012; Krumholz et al. 2012; Cunningham et al. 2018; Rosen & Krumholz 2020; Guszejnov et al. 2021; Mathew & Federrath 2021). As shown in Paper II, protostellar jets interrupt the accretion of material on to already-existing stars, while also promoting fragmentation and the birth of low-mass stars. This impact is clearly shown in the IMF of the star formation runs in Fig. 8. In comparison to their counterparts, star-forming runs that include protostellar jet feedback promote significantly higher quantities of low-mass stars, resulting in better agreement with the observed IMF.

The influence of protostellar jets upon the cloud kinematics is also heavily dependent on the star formation history, which varies

in the different setups. This is most clear when comparing the SFE and \mathcal{M} histories of the TURBSPHERE +Driving+Jets run to the BOX +Driving+Jets run, and their IMFs (Figs 7 and 8, respectively). While the IMFs of these two runs are qualitatively similar in shape, their final Mach numbers diverge. For the respective BOX run, the relative lack of stars results in far fewer protostellar jets and thus less energy injection into the system. While \mathcal{M} is sustained throughout the simulation (due to continual energy injection via our driving scheme), the smaller number of stars and their protostellar jets results in a reduced ability to stir turbulence, as evidenced by a stagnant \mathcal{M} in the BOX run. However, the more rapid star formation of the TURBSPHERE run allows for significant energy injection, allowing the Mach number to rise significantly. We can attribute this directly to the protostellar jets by comparing the TURBSPHERE +Driving run to the TURBSPHERE +Driving+Jets run, as the lack of protostellar jets results in an equilibrium Mach number being sustained, diverging from the run with protostellar jets enabled.

Lastly, although the IMF appears closer to the canonical Kroupa (2001) shape when protostellar jets are enabled in TURBSPHERE runs, there is still a residual excess of high-mass ($\gtrsim 10M_{\odot}$) stars seen in previous SPHERE simulations with protostellar jets (Paper II). However, these runs do not include other feedback mechanisms such as radiation pressure, photoionization, and stellar winds, which are likely important for limiting the accretion of massive stars.

4 DISCUSSION

4.1 Do initial conditions matter?

To determine the influence of initial conditions while holding boundary conditions fixed, we compare the decaying SPHERE and TURBSPHERE runs in Fig. 7, both with and without protostellar jet feedback. We note a delay in initial star formation in panel 1 for TURBSPHERE runs, where the SPHERE run begins forming stars approximately 1 Myr earlier. Despite this delay, the eventual star formation rate of TURBSPHERE is higher than that of SPHERE, resulting in TURBSPHERE overtaking SPHERE in SFE as time progresses. This is further supported in panel two, where the number of stars, N_* , diverges, favouring TURBSPHERE as time progresses. We attribute

these differences solely to the different initial conditions of the simulation.

As the TURBSPHERE simulation begins with self-consistent MHD turbulence, the magnetic field has already been amplified by the magnetic dynamo, and thus the gas initially has more magnetic pressure support, which has been shown to suppress star formation by reducing turbulent fragmentation (e.g. Rosen & Krumholz 2020; Appel et al. 2021). However, due to the realistic MHD turbulence of TURBSPHERE, pre-assembled turbulent structures, such as dense cores and filaments, are permitted to form *before* stars can form. Such cores provide pockets of dense gas where the Jeans mass is lower, encouraging fragmentation. As such structures form earlier in TURBSPHERE, star formation is encouraged at higher rates in comparison to SPHERE runs, once it begins. To a lesser degree, these phenomena (of delayed star formation and increased SFR in TURBSPHERE runs) are found when comparing TURBSPHERE+Jets and SPHERE+Jets runs.

To assess the impact of initial conditions on the qualitative structure of the IMF, we compare TURBSPHERE and SPHERE runs without jets or driving in Fig. 8. We find that the IMF turnover (where the power-law transitions to a flat segment) for the TURBSPHERE run occurs at slightly lower stellar masses than SPHERE runs. Compared to SPHERE, the TURBSPHERE setup produces higher quantities of low-mass stars for both runs with and without protostellar jets, however the difference in the shape of the IMF appears is much less pronounced with feedback.

Note that the IMFs of these runs were taken from the final snapshot of our simulations. As a result, the final SFE of our clouds differs between the plotted IMFs in Fig. 8 (final SFEs are listed in Table 2). We do not take the IMF at fixed SFE as BOX runs have a significantly suppressed SFE, making comparison to SPHERE and TURBSPHERE at fixed SFE difficult. Thus, we focus on the IMF turnover as our IMFs are likely more top-heavy than physically realistic. A more conclusive analysis exploring the origins of these differences requires a larger statistical sample of both SPHERE and TURBSPHERE setups with full-physics runs.

4.2 Disentangling the roles of driving and cloud boundary conditions

Simulations have historically varied both turbulence driving and boundary conditions together between the SPHERE and BOX setups, as a result it has been difficult to attribute various differences in star formation results to one factor or the other. Our numerical experiments allow us to disentangle these two effects.

4.2.1 Cloud boundary conditions

To isolate the role of periodic and non-periodic cloud boundary conditions on the results, we compare the BOX and TURBSPHERE +Driving runs in Fig. 7. Compared to the TURBSPHERE runs, we find significant reduction of the SFE and quantity of stars in BOX runs, with the SFE exhibiting a significantly shallower slope with respect to simulation time. This same SFE suppression has been noted previously between SPHERE and BOX setups in studies that have directly compared the two under controlled conditions (Krumholz et al. 2012; Guszejnov et al. 2020b) and has been broadly attributed simply to the moderation of star formation by turbulence (e.g. Krumholz & McKee 2005; Federrath & Klessen 2012). However, our results show that the cloud boundary conditions also contribute significantly.

To investigate the role of cloud boundary conditions on the IMF, we compare the results of TURBSPHERE and BOX runs in Fig. 8. We find that BOX produces relatively lower amounts of low-mass stars for runs with and without jets. We note that the IMF turnover skews towards higher masses in the BOX runs (as was also found in Paper II). We attribute the difference in the shift of the IMF turnover solely to the different cloud boundary conditions of TURBSPHERE and BOX.

Differences in both the SF history and stellar mass statistics are likely explained by the importance of global infall in the TURBSPHERE (and SPHERE) setups. As discussed in Section 1, for a given set of turbulent initial conditions the BOX setup generally has about an order of magnitude less gravitational energy due to the different boundary conditions of the Poisson equation. Consequently, BOX runs effectively have a much higher virial parameter (Federrath & Klessen 2012), inhibiting large-scale gravitational collapse. Conversely, both SPHERE and TURBSPHERE clouds are subject to their own global weight, and as a result tend to become centrally concentrated over time, despite driving. This may funnel gas to higher densities, resulting in more low-mass stars (due to a reduced Jeans mass) and encouraging star formation overall.

4.2.2 Effects of driving

The inclusion of turbulent driving in some of our simulations models large-scale energy injection produced by the larger galactic environment. To isolate the influence of driving, we primarily compare TURBSPHERE runs with and without driving enabled as we have a controlled experiment where non-driving factors are held fixed. We find that turbulent driving reduces the SFE and the number of stars at the same time interval. Lower SFEs and star counts are observed both with and without protostellar jets. We conclude that turbulent driving moderates star formation in its own right, as shown in previous works that aimed to disentangle various factors in a BOX-like setup (Federrath 2015).

Interestingly, driving in concert with protostellar jets appears to quench star formation in the cloud at an apparent final SFE approaching ~ 7 per cent. The SPHERE and undriven TURBSPHERE run with jets also show some evidence of the SFE tapering off (similar behaviour was also found in Paper II) but at significantly higher ($\gtrsim 15$ per cent) SFE. So in concert with feedback, driving can also reduce the final SFE of a GMC, presumably by providing pressure support (or in other words: increasing the virial parameter), which makes the cloud easier to disrupt. We note that the TURBSPHERE and SPHERE driving runs exhibit a top-heaviness in comparison to their non-driving counterparts in their IMFs. Since the growth of median and maximum stellar mass tends to correlate strongly with the SFE in simulations (Bonnell et al. 2003; Peters et al. 2010; Guszejnov et al. 2020b), earlier termination of star formation can be expected to moderate the top-heaviness of the IMF.

5 CONCLUSION

We have demonstrated a new setup for GMC-scale star formation simulations – called TURBSPHERE – that simultaneously models the self-consistent MHD turbulence, large-scale energy injection that realistically sustains large-scale turbulence, and spatial concentration of real GMCs. This new setup combines the advantages of previous SPHERE and BOX setups that are commonly used for simulating star formation that occurs within GMCs. Through the introduction of an analytic gravity well and continuous driving, we generated an isolated turbulent cloud in dynamical equilibrium (Figs 2 and 3). From our suite of turbulence driving runs, we find:

(i) The TURBSPHERE initial condition has properties resembling those found in the well-studied periodic BOX setup, as evidenced by the similarity in the density PDF and the velocity and magnetic field power spectra. We note some slight differences in the spectral indices and large-scale density structure, which we discuss in Section 3.1. (Figs 4, 5, and 6).

(ii) We map out the relations between the normalized forcing strength scale factor \tilde{f} , driving wavelength range scale factor $\tilde{\lambda}$, analytic gravity well scaling factor M_{well}/M_0 , and initial mass-to-flux ratio μ_0 for generating various equilibrium statistics (\mathcal{M}_{RMS} , R_{50} , R_{RMS} , $E_{\text{mag}}/E_{\text{kin}}$, and α_{turb}) for particular initial conditions generated by the TURBSPHERE setup. These power-law relations are listed in equations (6), (7), (8), (9), and (10), and discussed in Section 3.1.2.

We then performed a suite star formation simulations, with and without protostellar jets and driving, with various initial conditions (listed in Table 2). These results make it possible to isolate and study the effects of the turbulent cascade, simulation geometry, external driving, and gravity/MHD cloud boundary conditions upon the star formation history and IMFs predicted by simulations. We plot these quantities in Figs 7 and 8 and found:

(i) The periodic cloud boundary conditions of the BOX setup suppress gravitational collapse of the cloud, resulting in a significantly shallower SF history than the TURBSPHERE run with equivalent turbulent conditions and driving. The SF history of the TURBSPHERE simulations obeys a similar steep (SFE $\sim t^3$) law as previously seen in SPHERE simulations, regardless of driving or jet feedback. (Section 4.2.1)

(ii) Whether in the TURBSPHERE or BOX setup, and with or without feedback, the continued driving of turbulence moderates star formation significantly, with or without protostellar jets. Driven runs have an SFE that is a factor of ~ 2 – 3 lower at fixed times compared to their undriven counterparts. Driving in concert with jets can disrupt the GMC sufficiently to nearly halt star formation with an SFE $\lesssim 10$ per cent. (Section 4.2.2)

(iii) Without feedback, and with or without driving, the stellar mass function in TURBSPHERE runs remains much shallower and more top-heavy than observations, in agreement with previous conclusions from BOX and SPHERE runs (Guszejnov et al. 2020b). However, the detailed shape of the mass function does differ in each different setup, with TURBSPHERE having the lowest IMF peak mass of the three.

(iv) With protostellar jet feedback, TURBSPHERE runs produce far more low-mass stars and obtain a mass function more similar to e.g. the Kroupa (2001) form, again in agreement with previous results from the other setups (e.g. Paper II). However a shallower slope at very high masses ($>20M_\odot$) remains, indicating that additional physics are required to regulate the growth of the most massive stars.

We find that TURBSPHERE offers a more realistic approach to generating the initial conditions of isolated GMCs and modelling their subsequent gravitational collapse that drives star formation within them, however we note some caveats remain since GMCs are not isolated objects. It may be important to model the formation of GMCs within galactic contexts, notably through large-scale colliding flows, cloud–cloud collisions, and thermal and gravitational instability (Tasker & Tan 2009; Wu et al. 2017; Zamora-Avilés et al. 2019; Chevalier et al. 2020), which the TURBSPHERE setup does not model. Furthermore, while the difference between the analytic potential and the eventual gas self-gravity is small by construction, it is still unphysical to instantly ‘switch on’ self-gravity.

The TURBSPHERE setup permits simulations with turbulence driving in concert with all feedback mechanisms, including radiation, stellar winds, and supernovae, in addition to the protostellar jets considered here. Such simulations will allow a more in-depth study of the importance of large-scale energy injection on the star formation rate and the IMF in concert with feedback, bringing us closer to understanding star formation in GMCs in a galactic context. The results of these simulations and their analysis will be presented in future works.

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DATA AVAILABILITY

The data supporting the plots within this article and the initial conditions used for numerical tests are available upon request to the corresponding authors. A public version of the GIZMO code is available at <http://www.tapir.caltech.edu/~phopkins/Site/GIZMO.html>.

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SUPPORTING INFORMATION

Supplementary data are available at *MNRAS* online.

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APPENDIX A: EVOLUTIONS OF TURBSPHERE INITIALIZATION RUNS

In this appendix, we examine the turbulence statistics of the TURBSPHERE initialization runs listed in Table 1 and examined in Section 3.1. We plot the evolution of all runs listed in Table 1 and label them based on their altered parameter. Analysis of the macroscopic properties of each system, as shown in Fig. A1, reveals stability trends for all quantities of interest \mathcal{M}_{RMS} , R_{50} , R_{RMS} , $E_{\text{mag}}/E_{\text{kin}}$, and α_{turb} , as also seen in the fiducial run in Fig. 3 (equivalent to Panel A1a).

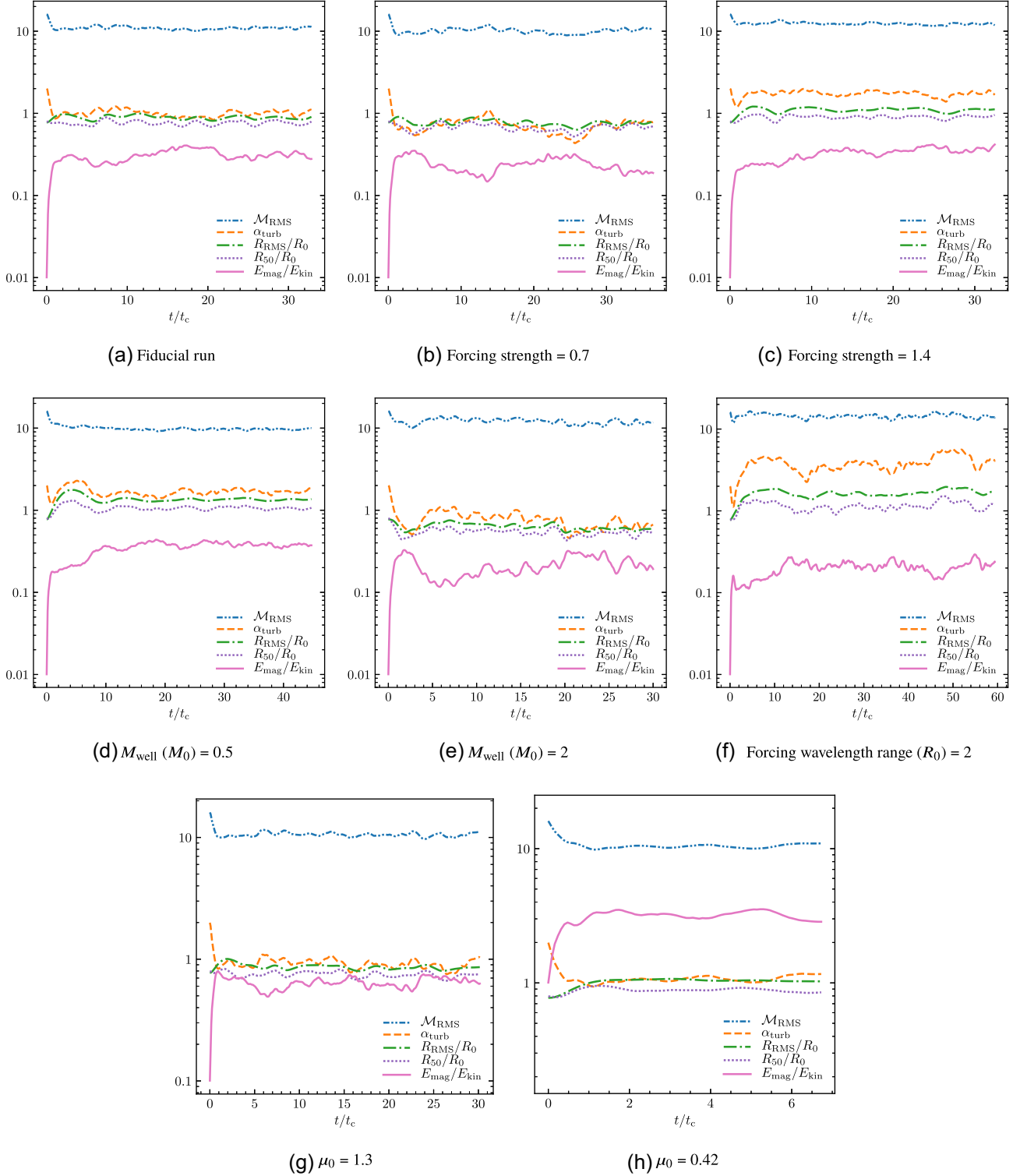


Figure A1. The evolutions of our TURBSPHERE initialization runs, labelled by their altered parameter. We plot the 3D RMS Mach number \mathcal{M}_{RMS} , half-mass radius R_{50} , RMS distance from centre of mass R_{RMS} , ratio of magnetic to turbulent energy $E_{\text{mag}}/E_{\text{turb}}$, and virial parameter α_{turb} as a function of time in units of the crossing time $t_{\text{cross}} = R_0/(c_s \mathcal{M}_{\text{RMS}})$. For all runs, all quantities approach equilibrium within ~ 10 crossing times.

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