Implicit large eddy simulations of anisotropic weakly compressible turbulence with application to core-collapse supernovae

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In the implicit large eddy simulation (ILES) paradigm, the dissipative nature of high-resolution shock-capturing schemes is exploited to provide an implicit model of turbulence. The ILES approach has been applied to different contexts, with varying degrees of success. It is the de-facto standard in many astrophysical simulations and in particular in studies of core-collapse supernovae (CCSN). Recent 3D simulations suggest that turbulence might play a crucial role in core-collapse supernova explosions, however the fidelity with which turbulence is simulated in these studies is unclear. Especially considering that the accuracy of ILES for the regime of interest in CCSN, weakly compressible and strongly anisotropic, has not been systematically assessed before. Anisotropy, in particular, could impact the dissipative properties of the flow and enhance the turbulent pressure in the radial direction, favouring the explosion. In this paper we assess the accuracy of ILES using numerical methods most commonly employed in computational astrophysics by means of a number of local simulations of driven, weakly compressible, anisotropic turbulence. Our simulations employ several different methods and span a wide range of resolutions. We report a detailed analysis of the way in which the turbulent cascade is influenced by the numerics. Our results suggest that anisotropy and compressibility in CCSN turbulence have little effect on the turbulent kinetic energy spectrum and a Kolmogorov $k^{-5/3}$ scaling is obtained in the inertial range. We find that, on the one hand, the kinetic energy dissipation rate at large scales is correctly captured even at relatively low resolutions, suggesting that very high effective Reynolds number can be achieved at the largest scales of the simulation. On the other hand, the dynamics at intermediate scales appears to be completely dominated by the so-called bottleneck effect, i.e., the pile up of kinetic energy close to the dissipation range due to the partial suppression of the energy dissipation rate by numerical viscosity. An inertial range is not recovered until the point where relatively high resolution $\sim 512^3$, which would be difficult to realize in global simulations, is reached. We discuss the consequences for CCSN simulations.

I. INTRODUCTION

Despite decades of studies and compelling evidence that a significant fraction of stars with initial masses in excess of $\sim 8$ solar masses explode as core-collapse supernovae (CCSN) at the end of their evolution, the exact details of the explosion mechanism are still uncertain\textsuperscript{2-5}. Current state-of-the art 3D simulations either fail to explode or have explosion energies that fall short of the observed energies by factors of a few for most of the progenitor mass range\textsuperscript{4,5,6}. The dynamics at the center of a star undergoing core collapse is shaped by a delicate balance between competing effects where all of the known forces: gravity, electromagnetism, weak and strong interactions, are important. The task of modeling these systems is made particularly challenging by the fact that the generation of the asymptotic explosion energies, although enormous ($\sim 10^{54}\text{J}$), requires a rather subtle, percent-level imbalance between nonlinear processes over many dynamical times.

The flow of plasma in the core of a star going supernova is known to be unstable to convection\textsuperscript{7-10} and/or to another large scale instability known as standing accretion shock instability\textsuperscript{11,12}. In any case, given the very large Reynolds numbers, as large as $\sim 10^{17}$ in the region of interest\textsuperscript{13} (the so-called gain region, where neutrino heating dominates over neutrino cooling), it is expected that the resulting flow will be fully turbulent. It has been suggested\textsuperscript{14,15} recently that turbulence and, in particular, turbulent pressure could tip the balance of the forces in favor of explosion. In this respect, anisotropy is of key importance, because it results in an effective radial pressure support with adiabatic index $\gamma_{\text{turb}} = 2$, much larger than that of thermal (radiation) pressure ($\gamma_{\text{th}} \simeq 4/3$). This means that turbulent kinetic energy is a much more valuable source of radial pressure support than thermal energy (see Appendix A).

All of the current numerical simulations employ the implicit large eddy simulation (ILES) paradigm\textsuperscript{16,17} (also known as monotone integrated LES (MILES)) of exploiting the dissipative nature of high resolution shock capturing (HRSC) methods as an implicit turbulence model. However, the combination of the use of rather dissipative schemes...
and the relatively low spatial resolution that can be achieved in global simulations is such that the fidelity with which turbulence is captured is questionable\textsuperscript{13}.

To be useful in the context of CCSN simulations, an ILES should, at the very least, account for the right rate of decay of the kinetic energy at the largest scales while avoiding unphysical pile up of energy at smaller scales. Unfortunately, all of the current simulations seem to be strongly dominated by the so-called bottleneck effect\textsuperscript{13}, which corresponds to an inefficient energy transfer across intermediate scales due to the viscous suppression of non-linear interaction with smaller scales\textsuperscript{18–22}. Current global simulations achieve resolutions, in the turbulent region, comparable to those of $30^3 - 70^3$ lattices in periodic domains\textsuperscript{13,15,23}. At these resolutions, almost all of the dynamical range of the simulations can be expected to be directly affected by numerical viscosity\textsuperscript{24}. The fidelity with which turbulence is captured in these simulations will then depend on the degree with which the numerical truncation error approximates an LES closure.

In this respect, it has been shown by Garnier et al.\textsuperscript{25} and Johnsen et al.\textsuperscript{26} that many HRSC methods can be too dissipative to yield a faithful description of turbulence at low resolutions. These studies, however, considered a different regime, decaying isotropic turbulence, while turbulence in a core-collapse supernova, as well as in many other astrophysical settings, is often strongly anisotropic\textsuperscript{14,15,27}, as rotational invariance is broken by gravity. Garnier et al.\textsuperscript{25} and Johnsen et al.\textsuperscript{26} also considered different numerical schemes with respect to those used in supernova simulations. Both of these aspects can, in principle, be important. First of all, strong anisotropies could potentially influence the turbulence dynamics at the level of the energy cascade and of the dissipation\textsuperscript{28}. Secondly, some of the schemes used in computational astrophysics, such as the piecewise parabolic method (PPM)\textsuperscript{29} as well as some of the MUSCL\textsuperscript{30} schemes, have been shown, differently from some of the methods considered by Garnier et al.\textsuperscript{25} and Johnsen et al.\textsuperscript{26}, to be well suited for ILES\textsuperscript{31,32}.

The aim of this work is to fill the gap between existing theoretical studies and the particular applications of our interest. To this end we use a publicly available code, FLASH, which is widely used in the computational astrophysics community, and perform a series of simulations of turbulence in a regime relevant for core-collapse supernovae: driven at large scale, with large anisotropies and mildly compressible. We use five different numerical setups and, for each, several resolutions in the range from $64^3$ to $512^3$ in a periodic domain. We study in detail the way in which the energy cascade across different scales is represented by our ILES and we discuss the use of local or lower dimensional diagnostics that can be used to assess the quality of a global simulation in a complex geometry where 3D spectra are not readily available.

The rest of this paper is organized as follows. First, in Section II, we discuss the exact setup of our simulations and the diagnostic quantities used in our analysis. Then, in Section III, we discuss the basic characteristics of the flow realized in our simulations. In Section IV, we present a detailed analysis of the way in which the energy cascade is captured by the different schemes at different scales. In particular, we quantify the accuracy with which different methods capture the decay rate of energy from the largest scales and the way in which energy is distributed across scales. We discuss the role of anisotropies in the context of the $4/5$–law, a fundamental exact relation for isotropic and incompressible turbulence relating the statistics of velocity fluctuations with the energy dissipation rate (see Section II C), in Section V. We explore the use of the 2D, transverse, energy spectrum as a diagnostic for 3D simulations in Section VI. Finally, we present a brief summary of our main findings, as well as a discussion of their implications for CCSN simulations in Section VII. Appendix A, contains some supplemental background material on the role of turbulence in the explosion mechanism of CCSN.

II. METHODS

A. Numerical methods

We consider a compressible fluid with a prescribed acceleration, $a$, in a unit-box with periodic boundary conditions. The code that we employ for these simulations, FLASH, solves the gas-dynamics equations in conservation form. In particular we evolve the continuity equation

$$\partial_t \rho + \nabla \cdot (\rho \mathbf{v}) = 0$$  \hspace{1cm} (1)

and the momentum equation

$$\partial_t (\rho \mathbf{v}) + \nabla \cdot (\rho \mathbf{v} \otimes \mathbf{v} + p \mathbf{I}) = \rho \mathbf{a}.$$  \hspace{1cm} (2)

These equations are closed with a simple isentropic equation of state,

$$p = \rho^{4/3},$$  \hspace{1cm} (3)
that can be considered as a rough description of a gas dominated by radiation pressure. Since the equation of state ensures an adiabatic evolution we do not need to solve the energy equation as equations (1), (2) and (3) suffice to fully describe the flow.

Equations (1) & (2) are solved using the directionally-unsplit hydrodynamics solver of the open-source FLASH simulation framework\textsuperscript{33–35}. FLASH implements the corner transport upwind method\textsuperscript{36} for fully directionally-unsplit evolution of the Euler equations\textsuperscript{37,38}. FLASH includes several options for the order of spatial reconstruction\textsuperscript{35}, including 2\textsuperscript{nd}-order TVD\textsuperscript{30}, 3\textsuperscript{rd}-order PPM\textsuperscript{29}, and 5\textsuperscript{th}-order WENOZ\textsuperscript{39}. Fluxes are computed at 2\textsuperscript{nd} order accuracy using one of a number of approximate Riemann solvers included in FLASH, such as HLLE\textsuperscript{40}, HLLC\textsuperscript{41}, and the Roe solver\textsuperscript{42}. Second-order accuracy in time is achieved via a characteristic tracing evolution of the Riemann solver input states to the time step midpoint\textsuperscript{29}.

Turbulence is driven using the stirring module of FLASH. This module uses the Ornstein-Uhlenbeck process\textsuperscript{43} to generate stirring modes in Fourier space. This yields an acceleration field which smoothly decorrelates\textsuperscript{44} over a timescale $T_s$. The FLASH implementation permits the use of any arbitrary combination of solenoidal and compressive modes\textsuperscript{45}. For our runs, we set $T_s = 0.5$, we use only solenoidal forcing and we restrict the accelerating field to be nonzero only in the first four Fourier modes. This forcing is designed to mimic the influence of some larger scale weakly compressible flow. In the CCSN context, simulations show that the turbulence is highly anisotropic, being roughly twice as strong in the radial direction as either tangential direction\textsuperscript{14,15,46,47} since it is driven by buoyancy due to a negative radial entropy gradient. In order to emulate this behavior, the accelerating field in the $x-$direction (which is going to play the role of the radial direction) is scaled by a constant factor (before the solenoidal projection of the acceleration field) such that $R_{xx} \simeq 2R_{yy} \simeq 2R_{zz}$, where

$$R_{ij} = \langle \rho v_i v_j \rangle,$$

is the Reynolds stress tensor (to simplify the notation we considered a frame in which $\langle \rho v \rangle = 0$) and $\langle \cdot \rangle$ denotes an ensemble average. Finally, the overall strength of the stirring is tuned to achieve a RMS Mach number of $\simeq 0.3$, which is typically observed in realistic CCSN simulations\textsuperscript{48,49}.

### B. Energy transfer equations

In order to study the cascade of the specific kinetic energy (which we will refer to simply as “kinetic energy” or “energy” in the following), $|\mathbf{v}|^2/2$, we will consider an energy budget equation across different scales, analogous to the one commonly employed in the study of incompressible, isotropic turbulence, e.g.,\textsuperscript{50}. In particular, we consider the momentum equation (2) in non-conservation form,

$$\partial_t \mathbf{v} + (\mathbf{v} \cdot \nabla) \mathbf{v} = -\nabla p + \mathbf{a},$$

where $\mathcal{V} = 1/\rho$ is the specific volume of the gas.

We can use equation (5) to derive an evolution equation for the Fourier transform of the velocity

$$\hat{\mathbf{v}}(\mathbf{k}) = \int_{\mathbb{R}^3} e^{-2\pi i \mathbf{k} \cdot \mathbf{x}} \mathbf{v}(\mathbf{x}) d^3\mathbf{x}.\tag{6}$$

Transforming both sides of equation (5) we obtain

$$\partial_t \hat{\mathbf{v}} - \hat{\mathbf{v}} * 2\pi i \mathbf{k} \otimes \hat{\mathbf{v}} = \hat{\mathbf{V}} * 2\pi i \mathbf{k} \hat{\mathbf{p}} + \hat{\mathbf{a}},$$

where $*$ denotes the convolution operator, i.e.,

$$[f * g](\mathbf{k}) = \int_{\mathbb{R}^3} f(\mathbf{q}) g(\mathbf{k} - \mathbf{q}) d^3\mathbf{q}.$$

If we multiply both sides of equation (7) by $\hat{\mathbf{v}}^*$ and take the real part, we obtain an equation for the 3D energy spectrum

$$\partial_t E(\mathbf{k}) = T(\mathbf{k}) + C(\mathbf{k}) + \epsilon(\mathbf{k}),$$

where

$$E(\mathbf{k}) = \frac{1}{2} \hat{\mathbf{v}} \cdot \hat{\mathbf{v}}^*,\tag{10}$$

$$T(\mathbf{k}) = 2\pi \Re (\hat{\mathbf{V}} * ik) \hat{\mathbf{p}} \cdot \hat{\mathbf{v}}^*,\tag{11}$$

$$C(\mathbf{k}) = 2\pi \Re (\hat{\mathbf{V}} * ik) \hat{\mathbf{p}} \cdot \hat{\mathbf{v}}^*,\tag{12}$$

$$\epsilon(\mathbf{k}) = \Re \hat{\mathbf{a}} \cdot \hat{\mathbf{v}}^*.\tag{13}$$
Here $E$ is the energy spectrum (the velocity power spectral density (PSD)) and $T$ is the same transfer term as in the classical incompressible equations and $\epsilon$ is the energy injection rate. The $C$ term vanishes in the incompressible limit and represents the interaction between kinetic and acoustic modes. In practice, in our models, $C$ is found to be at least one order of magnitude smaller than $T$ at all scales and it is thus negligible. In any case, we retain $C$ in the analysis below.

For each of the spectral quantities, $S$, being $E$, $T$, $C$ or $\epsilon$, we define the integrated spectrum, $S(k)$, as

$$S(k) = \int_{R^3} S(\mathbf{k}) \delta(|\mathbf{k}| - k) \, d\mathbf{k},$$  \hspace{1cm} (14)$$

$\delta(\cdot)$ being the Dirac delta function.

Integrating equation (9), we obtain the following one-dimensional energy balance equation

$$\partial_t E(k) = T(k) + C(k) + \epsilon(k).$$  \hspace{1cm} (15)$$

This can also be written in terms of the energy flux across scales,

$$\Pi(k) = -\int_{0}^{k} T(\xi) \, d\xi,$$  \hspace{1cm} (16)$$
as

$$\partial_t E(k) + \partial_k \Pi(k) = C(k) + \epsilon(k).$$  \hspace{1cm} (17)$$

Notice that we did not assume isotropy in any of the above.

Equation (15) is derived in the inviscid limit. In practice, our evolution method introduces dissipation in the form of “numerical viscosity”. This can be quantified in terms of the residual

$$R(k) = \partial_t E(k) - T(k) - C(k) - \epsilon(k).$$  \hspace{1cm} (18)$$

This can be used to define a wave number dependent “effective viscosity”:

$$\nu(k) = -\frac{1}{2} \frac{R(k)}{k^2 E(k)}. $$  \hspace{1cm} (19)$$

We remark that $\nu$ does not, in general, correspond to a classical shear or bulk viscosity, but can nevertheless be interpreted as a relative measure of the dissipation acting at different wave numbers (see, e.g., 51–53 for alternative approaches).

In practice, since we will be working in the stationary case, after having taken the appropriate time averages, $R(k)$ reduces to

$$R(k) = -T(k) - C(k) - \epsilon(k).$$  \hspace{1cm} (20)$$

Finally, since we are working in a periodic domain, which we take of size $L_x = L_y = L_z = 1$, all of the spectra are quantized and non-trivial only for $k_x, k_y$ and $k_z$ integers. Furthermore, all of the integrals in wave number space reduce to summations. Integrals over spherical shells are transformed to weighted sums following 44:

$$E(k) = 4\pi k^2 \sum_{k-1/2 < |\mathbf{k}| \leq k+1/2} E(k),$$  \hspace{1cm} (21)$$

where $N_k$ is the number of discrete wave-numbers in the shell $k - 1/2 < |\mathbf{k}| \leq k + 1/2$.

**C. Structure functions**

The energy spectrum and its sources/fluxes give a comprehensive picture of the energy cascade and can be used to assess the level of convergence of the simulation. Unfortunately, 3D energy spectra and fluxes are not easily accessible in calculations in complex domains and/or with inhomogeneous turbulence. In these cases, local quantities in the physical domain are more easily extracted and analyzed. Hence, one of the goals of this work is to validate
the use of indirect measures of convergence of ILES. Among these quantities, the structure functions of the velocity appear to be natural candidates for study.

We define the velocity increments
\[ \delta v(x, r) = \left[ v(x + r) - v(x) \right] \cdot \frac{r}{r} \]  
(22)
and study the quantities
\[ S_p(r) = \langle \delta v^p \rangle_{j=0} , \]  
(23)
where, \( \langle \cdot \rangle_{j=0} \) denotes an ensemble average as well as a mean over all of the angles between \( v \) and \( r \) (in other words we are looking at the \( j = 0 \) component of the SO(3) decomposition of the structure functions\(^{54}\)). In the case of homogeneous turbulence \( S_p \) does not depend on \( x \) and is thus a function of only the separation \( r \).

The most important relation involving the structure functions is the so-called \( 4/5 \)-law, which relates the third order structure function, \( S_3(r) \), with the mean energy dissipation rate,
\[ \langle \epsilon \rangle = \int_0^\infty \epsilon(k) \, dk , \]  
(24)
and states that, for incompressible, homogeneous and isotropic turbulence\(^{50}\):
\[ S_3(r) = -\frac{4}{5} \langle \epsilon \rangle \, r . \]  
(25)
Equation (25) can be derived from the Navier-Stokes equation for fully-developed, incompressible, homogeneous and isotropic turbulence\(^{50}\). In the anisotropic or compressible case, however, equation (25) is not strictly valid and could be violated in the data. As we show in Section V, we find equation (25) to be very well satisfied by our data, suggesting that the 3rd order structure function can be a very useful diagnostic in global simulations.

D. Transverse energy spectrum

Another alternative to the analysis of 3D spectra, which has been adopted by several authors in the core-collapse supernova context\(^{13,23,47,55}\), is the use of 2D spectra computed using a spherical harmonics expansion of the velocity field tangential to one or more spherical shells in the simulation. Analogously, we emulate this by looking at quantities in the \( y-z \) plane and we define the 2D spectra
\[ E_\perp(k_\perp) = \frac{1}{2} \int_{\mathbb{R}^2} \tilde{v}_\perp \cdot \tilde{v}_\perp^* \delta \left( \sqrt{k_y^2 + k_z^2} - k_\perp \right) \, dk_y \, dk_z , \]  
(26)
where \( \tilde{v}_\perp \) is the projection of the velocity perpendicular to the \( x \)-direction and we introduced the partial Fourier transform of \( v_\perp \):
\[ \tilde{v}_\perp(k_y, k_z) = \lim_{L_x \to +\infty} \frac{1}{L_x} \int_{-L_x/2}^{L_x/2} \int_{\mathbb{R}^2} e^{-2\pi i (k_y y + k_z z)} v_\perp(x, y, z) \, dy \, dz . \]  
(27)
In the limit of infinite Reynolds number / resolution, the 2D spectrum is expected to have the same asymptotic behavior as the 3D spectrum, however it is a-priori unclear if \( E_\perp \) is a good proxy for \( E \) at finite resolution. For this reason we find it useful to investigate this here.

As was the case for the 3D spectra, also here the spectrum is non-trivial only for integer \( k_y \) and \( k_z \), when periodicity is taken into account. The integral in equation (26) is treated analogously to the integral in the equation (14) for the 3D case, while the average in the \( x \)-direction in equation (27) is converted to an average over the \( x \)-extent of the simulation box.

III. BASIC FLOW PROPERTIES

We employ the finite-volume HRSC (Godunov) approach in which physical states are reconstructed at inter-cell boundaries and local Riemann problems are solved to compute the physical inter cell fluxes. In particular, we perform
For our fiducial quantities appear to have reached stationarity after time \( t \gtrsim 3 \) and oscillate around their target values. All resolutions produce the same qualitative behavior.

We assess how each numerical method performs when used under the same condition as in a real CCSN simulation where strong shocks need to be handled in some parts of the domain.

A first, qualitative comparison between the different methods can be done by looking at their visualizations. In particular, in Figure 2, we show a visualization of the magnitude of the vorticity in the \( x-z \) plane for four of the five schemes (excluding PPM_HLLC) at the highest resolution (512\(^3\)). The data is taken at the final time \((t = 100)\). As it can be seen from the figure, all of the simulations show the presence of thin, elongated, regions of high vorticity, as typically seen in direct numerical simulations (DNS) of homogeneous turbulent flows\(^{47,58}\). However, the width and the intensity of the vorticity at these smaller scales depend crucially on the numerical scheme. Methods with small intrinsic numerical viscosity, such as PPM_HLLC and WENOZ_HLLC, present smaller structures and more intermittent vorticity fields with respect to more dissipative methods, such as PPM_HLLC and TVD_HLLE.

Vortex tubes appear to be preferentially oriented along the \( x \)-axis, \( i.e., \) along the direction of anisotropy in the driving. Since regions of intense vorticity are also regions of intense dissipation for the classical Navier-Stokes equations, this anisotropy in the vorticity might result in an anisotropic dissipation tensor\(^{28}\). However, given the numerical nature of the viscosity that we employ and the relatively low coverage of the inertial range achieved in our simulations (see Section IV), we do not think that a quantitative analysis of this phenomenon is possible using our data. Furthermore, we also point out that in a CCSN anisotropy is not only due to the anisotropic driving, but also due to the fact that the rotational invariance of the hydrodynamics equations is broken by gravity. This will probably have an effect on the nature of vorticity structures at very small scales.

\[ 0 \leq t \leq 100 \]

\[ M_{\text{rms}} \]

\[ R_{\text{xx}}/R_{\text{yy}} \]

\[ R_{\text{xx}}/R_{\text{zz}} \]

\[ \text{PPM}_{\text{HLLC}}_{\text{N}64} \]

\[ \text{PPM}_{\text{HLLC}}_{\text{N}128} \]

\[ \text{PPM}_{\text{HLLC}}_{\text{N}256} \]

\[ \text{PPM}_{\text{HLLC}}_{\text{N}512} \]

\[ t \]

\[ 0 \leq t \leq 100 \]

\[ \text{PPM}_{\text{HLLC}}_{\text{CFL}0.8} \]

\[ \text{WENOZ}_{\text{HLLC}} \]

\[ \text{TVD} \]

\[ \text{HLLE} \]

\[ \text{Reynolds stress tensor} \]

\[ \text{RMS Mach number} \]

\[ \text{Reynolds stresses} \]

\[ (5 \gtrsim 3) \]

\[ \text{CCSN} \]

\[ \text{DNS} \]

\[ \text{Homogeneous turbulent flows} \]

\[ \text{Navier-Stokes} \]

\[ \text{Anisotropic dissipation tensor} \]

\[ \text{CCSN anisotropy} \]

\[ \text{Anisotropic driving} \]

\[ \text{Rotational invariance} \]

\[ \text{Hydrodynamics equations} \]

\[ \text{Gravity} \]
IV. THE ENERGY CASCADE

In this section we focus our analysis on the accuracy with which the energy cascade is captured by our ILES runs. First, we focus on the largest scales of the simulation with the goal of quantifying the accuracy in the decay rate of the energy as a function of the resolution for the different methods. Next, we will look at the energy distribution at smaller scales where, in resolved simulations, the inertial range starts. Finally, we will look at the dynamics in the dissipation region and summarize.
FIG. 3. Energy flux, as defined by equation (16), obtained with different numerical methods and resolutions. The energy flux is shown normalized to the average dissipation rate given by equation (24). From left to right and from top to bottom we show the results obtained with PPM_HLLC, PPM_HLLC_CFL0.8, PPM_HLLE, TVD_HLLE and WENOZ_HLLC. The bottom right panel show a comparison of all of the methods at 512\(^3\). All of the schemes show a good level of accuracy in the energy flux from the largest scales, with errors smaller than a few % already at low resolutions. The differences between the schemes become more marked at large wave numbers where the numerical dissipation starts to interfere with the energy cascade.

A. Energy decay rate

In the limit of very large Reynolds number it is assumed, in standard turbulence phenomenology\(^5^0\), that there exists a range of wave numbers (the inertial range) where energy injection and dissipation can be neglected in equation (17). In this range we can write (compressible effects are negligible in our simulations):

\[ \partial_t E(k) + \partial_k \Pi(k) \simeq 0, \]

so that stationarity requires \( \Pi(k) \simeq \text{const} \). In particular, since energy is conserved, one finds \( \Pi(k) \simeq \langle \epsilon \rangle \). This means that, in the limit of large Reynolds numbers, the energy decay rate depends only on the macroscopic properties of the flow (and in particular not on the nature of the viscosity), a fact that has also been verified numerically\(^5^9\). The significance of this property and its importance for the modeling of turbulence cannot be overstated.

In the context of CCSN simulations this means that the large scale kinetic energy, a crucial quantity for the dynamics of the explosion\(^1^5\), can be faithfully captured even with simulations achieving modest Reynolds numbers.

For an ILES, a basic requirement, then, is that a sufficiently high resolution should be achieved to correctly represent the energy cascade at the largest scales. What qualifies as a sufficiently high resolution is of course dependent on the details of the closure built into the scheme (and on the accuracy required for the particular application). To quantify this, we can estimate the level of accuracy that can be reached at any given resolution, using our local simulations. In particular, we can study directly the energy flux across scales, defined by equation (16). This is shown in Figure 3 for all of the different runs.

As discussed before, we expect that \( \Pi(k) \simeq \langle \epsilon \rangle \) over an extended region in Fourier space should be a direct indication that a simulation has been able to recover an inertial range. Perhaps not surprisingly, in light of previous results\(^2^4\), we find that regions where \( \Pi \simeq \langle \epsilon \rangle \) as wide as a few wave numbers \( 4 \lesssim k \lesssim 10 \) only appear at the highest resolutions (we will discuss the inertial range in more detail in Section IV B). However, the amount of energy decaying from the largest scales reaches an asymptotic value much quicker than that implying that the total kinetic energy budget at the largest scales is well resolved even at modest resolutions.
FIG. 4. Dissipation rate of the energy at the largest scales due to the turbulent cascade (not including direct dissipation by the numerical viscosity) as a function of resolution and for all of the schemes. The dissipation rate is normalized so as to be 1 in the limit of large Reynolds numbers / resolution. At $128^3$ points all of the schemes show an error of less than 10%, with the HLLC schemes already close to the 2% level.

We can make a more quantitative statement concerning the energy decay rate by looking at the peak of the energy flux as a function of resolution, as shown in Figure 4. We can see that at $128^3$ points all of the simulations have a deviation from the asymptotic energy decay rate of less than 10%. The least dissipative methods already have an error close to the 2% level. A comparison between PPM HLLE and PPM HLLC reveals the profound impact that the choice of the Riemann solver has even at relatively large scale (more on the dissipative properties of the different schemes in Section IV C).

B. Energy spectra

Obviously, not all of the dynamics of turbulence can be reduced to the rate at which kinetic energy decays from the injection scale. The internal dynamics of the energy cascade, far from the injection scale and far from the dissipation range, can also play an important role in many applications. To analyze this aspect we consider in Figure 5 the energy spectrum of the velocity defined by equation (10). The spectra are compensated by $k^{5/3}$ to highlight regions with Kolmogorov scaling, which might be expected in the inertial range. Since we want to focus on quantities that do not depend (or depend weakly) on the nature of the energy injection at large scale, we show all of the spectra as a function of a dimensionless wave number, $512 \frac{k}{\Delta x}$. The rationale behind this normalization is that, first of all, we assume the Kolmogorov scale $\eta$ to be proportional to the grid spacing. Secondly, the 512 factor is introduced to have the dimensionless $k$, $512 \frac{k}{\Delta x}$ coincide with the dimensional one for the highest resolution runs. With this choice, $512 \frac{k}{\Delta x} = 512$ corresponds to a wavelength of a single grid point, $512 \frac{k}{\Delta x} = 256$ corresponds to a wavelength of two grid points and so on.

Looking at any of the groups of runs in Figure 5, one can immediately notice that the spectra obtained at different resolutions do not collapse into a single curve in the dissipation region, as would be required by Kolmogorov’s first similarity hypothesis (cf., Gotoh et al. [60]). This lack of convergence in the dissipation region could be due to the non-linear viscosity of HRSC schemes. This, in turn, could result in an anomalous scaling of $\eta$ with the grid spacing. Such scaling has been reported in the past for ILES, but it is not very well understood [52]. The good agreement between the three different groups of simulations employing the HLLC Riemann solver seems to support this hypothesis and suggests that the nonlinear viscosity introduced by the Riemann solver is an important ingredient in setting this scaling.

Convergence appears to be recovered at larger scales $\gtrsim 8 \Delta x$ ($512 \frac{k}{\Delta x} \lesssim 64$), but the spectra appear to be dominated by the bottleneck effect. This manifests itself as a bump in the compensated spectra extending from the dissipation range until the end of the inertial range, for the simulations that show one (e.g., until $512 \frac{k}{\Delta x} = 10$ for the HLLC runs), or until the energy injection scale ($512 \frac{k}{\Delta x} = 4$), for the simulations that show no or little inertial range (TVD HLLE). The bottleneck effect is a viscous phenomenon which is also observed in direct numerical simulations. However, in the present context where viscosity is of numerical origin, it is at the very least questionable if a pronounced bottleneck is a desirable feature of the modeling. In astrophysical flows, where the Reynolds numbers...
FIG. 5. Energy spectra (equation 10) obtained with different numerical methods and resolutions. The energy spectra are compensated by a $k^{5/3}$ spectrum, so that any region with Kolmogorov scaling should appear roughly flat. Furthermore, the spectra are all plotted as a function of the dimensionless wave number $512 k \Delta x$ (the 512 factor is introduced to have the dimensionless wave number coincide with the dimensional one for the $512^3$ runs). The first five panels show the PPM HLLC (upper left), PPM HLLC CFL0.8 (upper center), PPM HLLE (upper right), TVD HLLE (lower left) and WENOZ HLLC (lower center) group of runs. The last panel (lower right) shows a comparison of all of the methods at the highest resolution ($512^3$). An inertial range seems to be recovered only at the highest resolutions (perhaps with the exception of TVD HLLE where no inertial range is visible). All schemes employing the HLLC Riemann solver are in very good agreement.

are typically very large, this pile up of energy at large scales is unphysical and could affect the quantitative and qualitative outcome of a simulation\(^\text{13}\). A quantification of the bottleneck effect in terms of the energy budget is discussed in Section IV D.

At even larger scales, an inertial range ($E \sim k^{-5/3}$ and $\Pi \sim \text{const}$, see Figure 3) seems to be recovered by the least dissipative schemes (PPM and WENOZ with HLLC) in the region $4 \lesssim k \lesssim 10$. PPM HLLE and TVD HLLE have a more limited region, a few wave numbers at most, that could be interpreted as being an inertial range. We note that this resolution is not particularly high in comparison with state of the art DNS\(^\text{59,61}\), but it would already correspond to an extremely high resolution in global CCSN simulations that typically have $\sim 30 - 70$ zones across the turbulent region\(^\text{13}\).

The overall behavior of the spectra, as obtained by all schemes, is consistent with Kolmogorov’s theory of turbulence. The anisotropic contributions to the angle-integrated spectra are too small to be detected in our data.

C. Effective viscosity

At very small scales ($\sim$ several grid points) the dynamics is dominated by the numerical viscosity. This can be estimated from the residual of the energy equation (17) or, equivalently, by the effective numerical viscosity $\nu(k)$ (equation 19). The latter is shown in Figure 6 for all of schemes and resolutions.

The first thing to notice is that the effective viscosity provided by all numerical schemes is not constant, but differs by roughly an order of magnitude between low and high $k$. Having a wave number dependent viscosity is a desirable feature expected in any LES model (explicit or otherwise). Nevertheless, this makes the definition and calculation of the Reynolds number achieved in a simulation ambiguous. Meaningful ways to estimate the Reynolds number from ILES have been proposed\(^\text{53}\) and they can be used to ease the comparison between different simulations and assess their quality. However, one has to be very careful while using any quoted Reynolds number from an ILES, to estimate...
FIG. 6. Effective viscosity as a function of the wave number measured for all schemes and resolutions. The effective viscosity is estimated using the procedure outlined in Section II and it is defined by equation (19). The different panels are, from left to right and from top to bottom, the results obtained with PPM_HLLC, PPM_HLLC_CFL0.8, PPM_HLLE, TVD_HLLE and WENOZ_HLLC. The bottom right panel show a comparison of all of the methods at 512^3. The numerical viscosity shows large variations across the wave number space. The choice of the Riemann solver plays a role that is at least as important as the choice of the reconstruction method in affecting the effective viscosity throughout the entire spectrum.

Two other features can be observed in most of the effective viscosity profiles. First, many of them exhibit a sudden reversal at high wave numbers. This is due to the fact that the numerical viscosity does not behave like a shear viscosity so that, although the numerical diffusion is strong at those scales, the effective viscosity appears small because of a partial decoupling between vorticity and dissipation. Second, at high resolution and at the largest scales, the effective viscosity is close to zero or even slightly negative. The reason is that the residual of equation (9) oscillates around zero and it is too small to be reliably extracted from our data: a much longer integration time would be needed to accumulate enough statistics for it.

Finally, a comparison between the effective viscosity reveals two interesting effects. First, by comparing PPM_HLLC and PPM_HLLE, we see that the choice of the Riemann solver affects the viscosity at basically all scales. Second, if we compare PPM_HLLC, PPM_HLLC_CFL0.8 and WENOZ_HLLC, we see that doubling the timestep has an effect comparable to changing the reconstruction method at intermediate scales (40 \lesssim k \lesssim 100).

D. The energy distribution

So far we have been concerned with the energy decay rate from the largest scales, which we have shown to be well captured by the ILES (Section IV A), and with the energy transfer in the inertial range, which we have seen to be described accurately only at much higher resolutions (Section IV B). In a turbulent flow both of these aspects are important and a good ILES should display a distribution of energy across vortical structures at different scales that is as close as possible to the asymptotic one. Obviously, there is a limit to the accuracy that any ILES can achieve at a fixed resolution. Here, we make this statement more quantitative by considering the amount of kinetic energy that is well resolved by each simulation at a given resolution.
FIG. 7. Cumulated energy distribution (equation 29) for all methods and resolutions, normalized by a factor $2/v_{rms}^2$ to be equal to 1 for large $k$. As a reference for comparison we also plot the asymptotic profile expected from Kolmogorov’s theory (equation 31). The different panels are, from left to right and from top to bottom, the results obtained with PPM_HLLC, PPM_HLLC_CFL0.8, PPM_HLLE, TVD_HLLE and WENOZ_HLLC. The bottom right panel show a comparison of all of the methods at $512^3$. At low resolution all of the schemes show an excess of energy at intermediate scales, due to the bottleneck. Only at the highest resolution at least, roughly, $80\%$ of the energy is correctly resolved.

We introduce the cumulative energy spectrum, the integral of the energy spectrum:

$$\mathcal{E}(k) = \int_0^k E(\xi) \, d\xi. \quad (29)$$

This quantity is plotted in Figure 7, where it is normalized by

$$\frac{v_{rms}^2}{2} = \int_0^{+\infty} E(k) \, dk \quad (30)$$

to obtain the cumulative distribution function of the kinetic energy. As a reference, we also show the cumulative energy spectrum estimated from Kolmogorov’s theory:

$$\mathcal{E}_{K41}(k) = \int_0^k E_{K41}(\xi) \, d\xi, \quad E_{K41}(k) = \begin{cases} E_{PPM\_HLLC\_N512}(k), & \text{if } k \leq 4, \\ E_{PPM\_HLLC\_N512}(4) \left(\frac{k}{4}\right)^{-5/3}, & \text{otherwise}. \end{cases} \quad (31)$$

We find that as the resolution increases, all schemes appear to be converging to the predictions of Kolmogorov’s theory. The results at finite resolution, however, are not encouraging: at $64^3$ only $\sim 50\%$ or less of the kinetic energy is in well resolved structures, while the other $\sim 50\%$ have piled up at rather large scale, with a cumulative excess of $\sim 20\%$ at the grid scale, mostly because of the bottleneck effect. At higher resolutions, the amount of kinetic energy well captured by the ILES increases, but at $512^3$ this is still only about $80\%$ of the energy and there is still a cumulative excess of over $\sim 5\%$ at the grid scale ($\ell \sim \Delta x$).

V. THE 4/5–LAW

The 4/5–law (equation 25) is not a-priori valid in the regime of turbulence we are considering. However, the 4/5–law has been numerically verified to hold also in some situations outside the domain of validity of its derivation.
FIG. 8. Compensated 3rd order structure functions (equation (23)) for all the numerical methods and resolutions. The structure functions are compensated and scaled so that they should be close to one where the 4/5-law (equation (25)) is verified. The data is plotted as a function of the dimensionless separation \( r/\Delta x \). The first five panels show the PPM_HLLC (upper left), PPM_HLLC_CFL0.8 (upper center), PPM_HLLE (upper right), TVD_HLLE (lower left) and WENOZ_HLLC (lower center) group of runs. The last panel (lower right) shows a comparison of all of the methods at the highest resolution \((512^3)\). The 4/5-law is very well verified in our data suggesting that 1) anisotropic corrections are subdominant and 2) all of the simulations achieve very large Reynolds numbers on the largest scales.

For instance, for isotropic mildly compressible decaying\(^{62}\) and driven\(^{63}\) turbulence. In the anisotropic case, however, anisotropic contributions cannot be excluded\(^{64}\), although they are known to be subdominant in some important cases\(^{65-67}\). In this section we show that equation (25) is consistent with our data over a wide range of scales.

We compute the 3rd order structure functions of the velocity, defined by equation (23), in a rather simple way using a random sample of 20,000 points in each of the 380 3D data dumps of our simulations. At each time, we compute the 3rd power of the velocity increments for each pair of points and accumulate and average in time the results in bins of size \( \Delta x \). The resulting structure functions are shown in Figure 8, compensated by \( -\frac{5}{r} (\langle \epsilon \rangle)^{-1} (\delta v^3)_{j=0} \), so that the resulting quantity should be equal to one if the 4/5-law is satisfied in our data. As was the case for the energy spectra, we assume \( \eta \sim \Delta x \) and plot the structure functions versus \( r/\Delta x \).

The degree with which the 4/5-law is satisfied in our data is very good. We see that anisotropic contributions only play a minor role in the angle-integrated formulation of the 4/5-law. This is in agreement with the incompressible DNS of Kaneda et al.\(^{67}\) and has been known to be true also for Rayleigh-Bénard convection in most regimes\(^{68}\). Our results provide an important new example where this appears to hold true. Secondly, for all of our simulations at \(512^3\), we find

\[
\max_r \left\{ -\frac{5}{4} r^{-1} (\langle \epsilon \rangle)^{-1} (\delta v^3)_{j=0} \right\}
\]

within 5% of 1. This level of accuracy is reached in DNS simulations achieving at least a Taylor micro-scale Reynolds number\(^{67}\)

\[
R_\lambda = \frac{u' \lambda}{\nu} \sim 300 ,
\]

where \( \nu \) is the kinematic viscosity, \( u' = \sqrt{\lambda} \nu_{rms} \) and \( \lambda = (15 \nu u'^2 / \langle \epsilon \rangle)^{1/2} \) is the Taylor micro-scale. This corresponds to a large-scale Reynolds number \( R = \frac{u' L}{\nu} \sim R_\lambda^2 \), \( L = 1 \) being the domain size, in excess of \( \sim 85,000 \). Reaching
FIG. 9. 3D (equation 10, blue) and 2D, transverse (equation 26, red) energy spectra for the PPM\_HLLC simulations. The energy spectra are compensated by $k^{5/3}$ to highlight eventual regions with Kolmogorov scaling. The spectra are plotted as a function of the dimensionless wave number $512 k \Delta x$, as in Figure 5. Although $E(k)$ and $E_{\perp}(k_{\perp})$ have similar trends, the use of the transverse spectrum can overestimate the width of the bottleneck region.

these Reynolds numbers in a DNS requires resolutions between $512^3$ and $1024^3$ using pseudo-spectral methods. This large-scale estimate of the Reynolds number is consistent with previous findings, although it is several orders of magnitude larger than the one that can be estimated using $\nu_{\text{max}}$. For instance, for PPM\_HLLC at $512^3$, $\nu_{\text{max}} \simeq 1.5 \times 10^{-3}$ and $v_{\text{rms}} \simeq 0.4$ giving

$$\frac{u' L}{\nu_{\text{max}}} \simeq 150.$$ \hspace{1cm} (34)

This apparent discrepancy is due to the fact that an ILES is not a DNS and, in particular, there is no unique Reynolds number for an ILES. As a consequence, different quantities that in a DNS depend on the Reynolds number, such as the dissipation rate or the Kolmogorov scale, behave as though the simulation has multiple values of the Reynolds number.

VI. THE TRANSVERSE SPECTRUM

Finally, we want to comment on the use of 2D transverse spectra in 3D simulations, a practice typically employed in the analysis of turbulence in CCSN simulations. Figure 9 shows a comparison of the 2D transverse spectrum $E_{\perp}(k_{\perp})$ from equation (26) and the 3D energy spectrum from equation (10) for the PPM\_HLLC simulations. The other runs (not shown) have the same qualitative behavior. We can see that the transverse spectrum follows qualitatively the same trend as the 3D spectrum in terms of convergence. They are both roughly compatible with a Kolmogorov scaling, but the bottleneck appears to be more pronounced in the 2D spectrum than in the 3D spectrum. In particular, $E_{\perp}(k_{\perp})$ only shows a very small region that suggests an inertial range, $3 \lesssim k \lesssim 5$ (as opposed to $5 \lesssim k \lesssim 10$ in $E(k)$).

Abdikamalov et al. concluded, also based on the analysis of 2D spectra, that turbulence in CCSN simulations is dominated by the bottleneck effect. Given the resolutions used in CCSN studies, our work supports their conclusion. However, in the light of Figure 9, we recommend that future studies supplement the analysis of 2D spectra with 3rd order structure functions, that, as we have shown, can give a more accurate description of the energy cascade.

VII. CONCLUSIONS

The details of the explosion mechanism of CCSNe have eluded our comprehension in spite of more than 50 years of studies. Recent numerical advances suggest that turbulence might play a fundamental role in tipping over the balance of the forces and lead to successful explosions (see also Appendix A). At the same time, the level of
accuracy of current simulations, which employ the ILES methodology, is unclear\textsuperscript{13}. Turbulence in CCSNe is mildly compressible, but strongly anisotropic\textsuperscript{4,13}. Simulations use rather dissipative numerical schemes (because they have to deal with strong shock waves and complex microphysics) and relatively low resolution, a combination (anisotropic turbulence and dissipative schemes) that has not been systematically studied before.

With the goal of assessing the reliability of ILES employed in the study of CCSNe, as well as in other areas of physics and astrophysics, we performed a series of local simulations of driven, anisotropic, weakly compressible turbulence. We compared five commonly employed numerical schemes with different reconstruction methods, Riemann solvers, and time step size. Each was run at 4 different resolutions ranging from $64^3$ to $512^3$. Our analysis focused on the fidelity with which the turbulent cascade is represented in each model. In particular, we performed an analysis both in Fourier space (with the velocity power-spectra and the energy flux) and in physical space (with the 3\textsuperscript{rd} order structure functions). Finally, we measured the effective viscosity of each scheme from the residual of the specific kinetic energy equation.

We found that, on the one hand, all of the numerical setups are able to accurately capture the decay rate of kinetic energy from the injection scale, with errors at the few \% level already at $128^3$ (e.g., $\sim 2.5\%$ for $\text{PPM}_{\text{HLLC}}N_{128}$). On the other hand, a large fraction of the energy is at unresolved scales where it piles up due to the bottleneck effect and an inertial range appears only at the highest resolutions ($512^3$). Even at this resolution, which would be difficult to achieve in global simulations, only roughly $\sim 80\%$ (the exact number depends on the scheme, see Section IV D) of the energy is resolved, the remaining $\sim 20\%$ accumulates as excess energy at intermediate scales (the cumulative energy excess at the grid scale alone is as large as $\sim 5\%$ of the total energy).

Current CCSN simulations have resolutions of at most 30 – 70 points covering the gain region\textsuperscript{13} (the energy injection scale). Based on our analysis we expect that at these resolutions even the energy decay rate from the largest scales will not be completely converged, but will show errors of up to tens of percent, depending on the numerical scheme (see Section IV A). At smaller scales, the dynamics is going to be completely dominated by the bottleneck effect. This is in agreement with the findings of Abdikamalov \textit{et al.}\textsuperscript{13}, based on the use of global simulations reaching a maximum resolution of 66 grid points covering radially the extent of the gain region.

Based on our findings, we expect that, if the resolution in global simulations is increased by a factor $\sim 2$ from the one of Abdikamalov \textit{et al.}\textsuperscript{13}, the decay rate will be converged to within a few \% of the asymptotic value. This implies that the ratio between thermal and kinetic energy, a crucial quantity for the onset of the explosion, will also be converged to within a few \%, at least when the energy injection rate changes slowly compared to the eddy turnover time (which is roughly $\sim 20$ ms in a CCSN\textsuperscript{23,70}). Unfortunately, while the lead up to explosion occurs over a larger timescale of a few hundred milliseconds, the transition to explosion can happen over much shorter timescales (one turnover time or less)\textsuperscript{13}. This means that the dynamics of the cascade over smaller time and length scales in the gain region also needs to be captured correctly since changes in the energy input rate on such short time scales will yield an inaccurate representation of the energy on large scales do to the bottleneck effect. This could require an increase of resolution by a factor $\sim 4 – 8$ with respect to current high-resolution simulations. Additional work using semi-global or global simulations will be required to more firmly establish the resolutions requirements at the transition of the explosion.

Concerning the properties of anisotropic turbulence in our simulations, we found anisotropy contributions to the energy spectrum and to the angle-averaged formulation of the $4/5$–law to be subdominant: the accuracy with which the $4/5$–law is satisfied is limited only by the employed resolution and the energy spectrum appears to be consistent with Kolmogorov $k^{-5/3}$ scaling. We also found the transverse energy spectrum with respect to the direction of anisotropy, a quantity typically computed in CCSN simulations, to overestimate the bottleneck with respect to the angle-integrated 3D spectrum. For this reason, we recommend future studies of CCSN to supplement (or replace) the analysis of the transverse spectrum with the analysis of the 3\textsuperscript{rd} order, angle-integrated, structure function (or, where possible, with the 3D spectrum itself).

Our results are, of course, dependent on the choice of the numerical scheme. In particular, we found significant differences in the dissipative properties of schemes employing the HLLC Riemann solver with respect to schemes using the more dissipative HLLE solver. The reconstruction order of the scheme is also important, although, while significant differences are found between TVD and PPM, the differences between PPM and WENOZ are much more minute (despite WENOZ being significantly more computationally expensive than PPM). In the end, none of the schemes we considered seems to be able to yield an accurate representation of the kinetic energy distribution across different scales at an affordable resolution for global CCSN simulations. A possible way forward would be to adopt low-dissipation numerical schemes especially designed for the use in ILES, such as the methods proposed by Hickel \textit{et al.}\textsuperscript{71}, Martin \textit{et al.}\textsuperscript{72} or Thornber \textit{et al.}\textsuperscript{73}. Implementing and testing these schemes will be subject of future work.

An important limitation of the present work is that we considered a very idealized setup. On the one hand, this allows us to benchmark the behavior of ILES in a controlled environment. On the other hand, our simulations cannot fully capture all features of the turbulent convective flow in a CCSN. Unlike the situation in a CCSN, our local simulations did not include a vertical advective velocity field that is due to the accretion of the stellar mantle. However,
the advective velocities are nearly constant in the regions of interest and Galilean invariance ensures that our results are unaffected. More limiting is the local nature of our simulations and the inevitable choice of boundary conditions. Moreover, our simulations could not take into account spatial variations in gravity and the large-scale radial convergence of the flow in globally spherical problems like collapsing stars. Addressing these issues will also be subject of future work.

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Appendix A: The role of turbulence in core-collapse supernova explosions

In this appendix we present a brief discussion of the importance of turbulent pressure in the explosion of massive stars. To set the stage, we will briefly summarize what is known of the dynamics of the most common class of CCSNe that are relevant for our later discussion. This is done for the benefit of readers that are not supernova specialists and it is not meant to be a complete review of the status of the field, for which we refer, instead, to the reviews of Janka et al.\textsuperscript{3} and Burrows\textsuperscript{4}. Next, we will discuss the role of turbulence and, in particular, of turbulent pressure on the explosion mechanism, in light of some recent results\textsuperscript{14,15}.

1. The neutrino mechanism

Towards the end of their evolution, massive stars form massive ($\sim 1.5$ solar mass) iron cores at their center. Since the iron nucleus has the largest binding energy per nucleon, no energy can be extracted from nuclear fusion beyond iron. The iron core is essentially inert and supported against gravity only by the degeneracy pressure of relativistic electrons. The mass of the iron core increases with time as more iron-group material is added by silicon shell burning. Electron capture on protons, which becomes energetically favorable at high densities, depletes the core of electrons and thus reduces the pressure supporting it against gravity. Eventually, the core becomes dynamically unstable and collapses.

During the collapse, the subsonically collapsing inner core ($\sim 0.5$ solar masses) contracts until it reaches densities comparable to that in atomic nuclei ($\sim 4 - 5 \times 10^{14}$ g cm$^{-3}$). At this point, the nuclear equation of state stiffens (due to the strong nuclear force). This halts the collapse of the inner core. It stops, bounces back and a proto-neutron star (PNS) is formed. The outer core, however, is still collapsing supersonically and a strong shock wave is launched at the interface between the inner and outer core.

It was once thought that this shock wave would travel outwards dynamically, depositing its energy in the outer layers of the star, causing the explosion. However, multiple numerical simulations performed over multiple decades have consistently shown that the initial shock fails to explode the star. Instead, it stalls due to energy losses to the dissociation of heavy nuclei into free nucleons and to the emission of neutrinos that stream away from the neutrino-semitransparent regions behind the shock.\textsuperscript{74} The shock generally stalls within only a few tens of milliseconds of core bounce and turns into an accretion shock standing at a radius of $\sim 100 - 200$ km. The accretion rate through the shock is so high (a fraction of a solar mass per second) that, if nothing revitalizes the shock within $\sim 1 - 2$ seconds, the gravitational force would overwhelm the nuclear repulsion force, collapsing the core of the supernova to a black hole, precluding explosion (e.g.,\textsuperscript{75}).

During this time, however, the PNS will release a significant fraction of its binding energy in the form of neutrinos (of order $10^{50}$ J). Converting a few percent of that energy into kinetic energy would be enough to unbind the stellar envelope and power the supernova explosion. In the standard neutrino mechanism it is theorized that a small fraction neutrinos emitted from the edge of the protoneutron star is re-absorbed in the region right behind the stalled shock. The deposition of neutrino energy leads to higher thermal pressure so that the shock can eventually overcome the ram pressure of accretion and accelerates in a run-away process\textsuperscript{74,76}. Turbulence in the heating region behind the shock
increases the time a fluid parcel spends in that region and, importantly, turbulent pressure helps in overcoming the ram pressure of accretion (see next Section and\(^2\)). It is, however, presently unclear if neutrino heating (even if aided by turbulence in launching the explosion) is able to provide enough energy to power the explosions to the energies inferred from astronomical observations.

2. Turbulent pressure and the Rankine-Hugoniot conditions

Simulations\(^3,14,15\) have shown that turbulence and, in particular, turbulent pressure behind the shock, could play an important role in aiding the explosion. To see why this is the case, we consider the Rankine-Hugoniot momentum condition for a standing accretion shock in a supernova core,

\[
s[\rho_d v_d^r - \rho_u v_u^r] = \rho_d (v_d^r)^2 + p_d - \rho_u (v_u^r)^2 - p_u, \tag{A1}
\]

where \(s\) is the shock speed and \(v_d\) and \(v_u\) denote the downstream and upstream values respectively. For the purpose of our discussion, we can assume the upstream gas to be cold and free-falling:

\[
p_u = 0 \quad (v_u^r)^2 = \frac{GM}{r}. \tag{A2}
\]

For the shock to expand we must then have

\[
\rho_d (v_d^r)^2 + p_d > \rho_u \frac{GM}{r}. \tag{A3}
\]

In the presence of turbulence, Murphy et al.\(^14\) suggested to modify equation (A3) in a way akin to a Reynolds decomposition and write it as

\[
\rho_d (\bar{v}_d^r)^2 + \rho_d (\delta v_d^r)^2 + p_d > \rho_u \frac{GM}{r}, \tag{A4}
\]

where \(v\) is the average velocity and \(\delta v = \bar{v} - v\) is the turbulent velocity. Although not entirely rigorous, equation (A4) has been shown to be well verified in the numerical simulations if angular averages are used to compute the respective quantities.\(^14,15\) Couch and Ott\(^15\) have shown that the turbulent pressure expressed in this fashion can exceed 50% of the thermal pressure, making a very significant contribution to the momentum balance in (A4).

Going beyond the arguments of Murphy et al.\(^14\), we can reinterpret equation (A4) as being the Rankine-Hugoniot condition for a fluid with a modified equation of state, which has two separate internal degrees of freedom: thermodynamical and turbulent. To this aim, we express \(\delta v_d^r\) in terms of the specific turbulent energy

\[
e_{\text{turb}} = \frac{1}{2} |\delta v|^2 \tag{A5}
\]

noting that

\[
|\delta v|^2 := (\delta v^r)^2 + (\delta v^\theta)^2 + (\delta v^\phi)^2, \tag{A6}
\]

and using the fact that

\[
(\delta v^r)^2 \simeq (\delta v^\theta)^2 + (\delta v^\phi)^2 \tag{A7}
\]

in CCSN turbulence, to obtain

\[
(\delta v^r)^2 \simeq \frac{1}{2} |\delta v|^2 = e_{\text{turb}}. \tag{A8}
\]

Assuming the pressure varies like \(p = (\gamma - 1)\rho e\), and substituting (A8) into (A4), we find

\[
\rho_d (\bar{v}_d^r)^2 + (\gamma_{\text{th}} - 1) \rho_d e_d + (\gamma_{\text{turb}} - 1) \rho_d e_{\text{turb}} > \rho_u \frac{GM}{r}, \tag{A9}
\]

where \(\gamma_{\text{th}} \simeq 4/3\) is the thermodynamical adiabatic index, \(e_d\) is the downstream thermal energy and \(\gamma_{\text{turb}} = 2\) is the equivalent adiabatic index associated with anisotropic CCSN turbulence. Since \(\gamma_{\text{turb}} > \gamma_{\text{th}}\), we see that turbulent energy is more efficient, per unit specific internal energy, at pushing the shock than thermal energy.
We point out that, if equation \((\text{A7})\) is dropped and turbulence is assumed to be isotropic, then \(\gamma_{turb} = 5/3\), which is still larger than \(\gamma_{th}\), but not as large as for the anisotropic case. This is a simple consequence of the fact that anisotropic turbulence has an anisotropic pressure, which is stronger in the radial direction.

In both cases, since the total energy is conserved, the relevant quantity is the ratio \(\epsilon_{turb}/e\). From standard turbulent phenomenology we expect that this ratio will only depend on macroscopic parameters, such as the net heating rate, the accretion rate and so on, and not on the details of the viscosity. For this reason, we expect this ratio to be correctly captured in ILES achieving a sufficiently high resolution.

As a final remark, we point out that a similar argument has been recently proposed by Mueller and Janka who formulated their equations in terms of the turbulent Mach number, as opposed to the turbulent energy.

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