

Evolution of semilocal string networks. II. Velocity estimatorsA. Lopez-Eiguren,^{1,*} J. Urrestilla,^{1,†} A. Achúcarro,^{2,1,‡} A. Avgoustidis,^{3,§} and C. J. A. P. Martins^{4,5,||}¹*Department of Theoretical Physics, University of the Basque Country UPV-EHU, 48040 Bilbao, Spain*²*Institute Lorentz of Theoretical Physics, University of Leiden, 2333CA Leiden, Netherlands*³*School of Physics and Astronomy, University of Nottingham, University Park, Nottingham NG7 2RD, England*⁴*Centro de Astrofísica da Universidade do Porto, Rua das Estrelas, 4150-762 Porto, Portugal*⁵*Instituto de Astrofísica e Ciências do Espaço, CAUP, Rua das Estrelas, 4150-762 Porto, Portugal*

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We continue a comprehensive numerical study of semilocal string networks and their cosmological evolution. These can be thought of as hybrid networks comprised of (nontopological) string segments, whose core structure is similar to that of Abelian Higgs vortices, and whose ends have long-range interactions and behavior similar to that of global monopoles. Our study provides further evidence of a linear scaling regime, already reported in previous studies, for the typical length scale and velocity of the network. We introduce a new algorithm to identify the position of the segment cores. This allows us to determine the length and velocity of each individual segment and follow their evolution in time. We study the statistical distribution of segment lengths and velocities for radiation- and matter-dominated evolution in the regime where the strings are stable. Our segment detection algorithm gives higher length values than previous studies based on indirect detection methods. The statistical distribution shows no evidence of (anti)correlation between the speed and the length of the segments.

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Topological defect networks [1–4] are predicted to form in high-energy physics models of the early universe [3,5,6]. In many cases they are stable enough to survive through the subsequent expansion history of the universe, as fossil relics of its earlier stages. Advances in observational astrophysics and cosmology including sensitive observations of the cosmic microwave background [7–12], pulsar timing arrays [13,14] and more recently the detection of gravitational waves [15,16] provide exciting opportunities to characterize or at least constrain the physics of the early universe through cosmic strings.

Among the various types of possible defect networks, string-type ones are the most common (their formation being inevitable in many models) as well as the most interesting since (unlike other defects, such as monopoles and domain walls) their evolution is broadly speaking cosmologically benign. Defects of this type include the simplest Nielsen-Olesen line-like topological defects in field theories breaking $U(1)$ symmetry [17], coherent macroscopic states of fundamental superstrings (known as F-strings), and D-branes extended in one macroscopic direction (D-strings). The latter two examples, collectively referred to as cosmic superstrings [18], are generically

predicted in string theoretic inflationary models involving spacetime-wrapping D-branes [6,19–21]. Various other hybrid networks are also possible, including the semilocal strings [22,23] that are the focus of this work.

Understanding the evolution of these defect networks, and in particular the distinguishing features of each specific model class, is essential in order to fully exploit the forthcoming data. Since these are intrinsically nonlinear objects, the quantitatively accurate modeling of defect network evolution is a difficult problem: analytic models can be obtained (using reasonable approximations), but any such model will necessarily contain some free parameters which cannot be calculated *ab initio*, but must be inferred by direct comparison with sufficiently high resolution numerical simulations—thereby calibrating the analytic model. The canonical approach to this modeling process is embodied in the velocity-dependent one-scale model (hereafter VOS model) and its various extensions [24–29]. Naturally, the complexity of the model is directly related to the complexity of the underlying particle physics model.

In the simplest scenario, corresponding to the Nielsen-Olesen solutions of the Abelian Higgs (hereafter AH) model there is clear analytic and numerical evidence that in simple radiation or matter dominated universes the network will reach a self-similar scaling regime, characterized by a single length-scale—the correlation length—which asymptotes to a constant fraction of the horizon. This allows us to use numerical simulations of limited dynamical range to infer properties of string networks evolving in the real Universe over much larger time scales. In practice, in a realistic

*asier.lopez@ehu.es

†jon.urrestilla@ehu.es

‡achucar@lorentz.leidenuniv.nl

§anastasios.avgoustidis@nottingham.ac.uk

||Carlos.Martins@astro.up.pt

universe containing both radiation, matter and dark energy, a cosmic string network never reaches this asymptotic scaling regime [30], but it approaches it relatively fast. The situation is less clear for non-Abelian strings, as well as for cosmic superstring networks, despite much recent activity in this area.

Semilocal strings are a hybrid type of defect arising in theories with both local and global symmetries [22,23]. For example the standard semilocal model is a minimal extension of the Abelian Higgs model by a global $SU(2)$ symmetry. This model, which has an $SU(2)$ doublet of two equally charged Higgs fields under a single $U(1)$ gauge field, admits stable string solutions even though the vacuum manifold is simply connected. They are also well-motivated from the theoretical point of view, arising in supersymmetric grand unified theories of inflation [31] and the corresponding D3/D7 brane inflation models [32]. These are a natural extension of usual inflationary models, in which the only extra ingredient is the doubling of a hypermultiplet. Being nontopological, semilocal strings possess remarkably different microphysical properties, as compared to their topological counterparts. The most noteworthy of these is that they form finite open segments whose ends have long-range interactions akin to global monopoles [33].

The cosmological evolution of semilocal networks is correspondingly distinctive. Their dynamics is dominated by the long range forces between the monopoles which imply that the individual segments can shrink and annihilate or grow by joining with other segments. The analytic modeling of these networks in a VOS-like context was started in [34], based on the premise that they can be described as local strings ending on global monopoles.¹ In addition to describing the overall evolution of the network, this also includes a phenomenological description of the evolution of individual semilocal segments and a preliminary comparison to numerical simulations, though all these were somewhat limited by the lack of numerical simulations with adequate resolution.

More recently, we have taken advantage of progress in computing power and started a more systematic study of the cosmological evolution of semilocal string networks, by presenting the first detailed numerical study of these networks [36], exploring a broad range of relevant cosmological and particle physics parameters and focusing on their large-scale properties. We have found evidence for scaling behavior of semilocal networks, at least in the sense that both the total string length of the network and the number of string segments evolve towards linear scaling at late times, and we have demonstrated consistency with the predictions of the aforementioned VOS model.

The main bottleneck in previous works, as well as the main source of uncertainties in our modeling—indeed

¹The VOS model for global monopoles has been recently revisited in [35].

preventing a fully quantitative calibration—is our ability to measure the transverse velocity of the string segments and the longitudinal velocities of the monopoles in our simulations. Addressing this issue required the introduction and extensive testing of various numerical techniques. The goal of this paper is to describe these techniques, discuss their validation, and present the key results of the analysis of the string and monopole velocities. It will be seen that our results provide supporting evidence for the scaling of these networks, while also providing additional insights into the physical mechanisms underlying the dynamics of these networks.

A novel aspect of this work is the introduction of an algorithm that identifies the position of the cores of semilocal string segments much more precisely than in previous works. This allows a more accurate determination of the length and also of the velocity of *each* segment and how these evolve in time, as well as their averages and distribution across the network. It has revealed some interesting aspects of the velocity distribution, in particular, that had not been reported previously. We return to this point in the conclusions. In a subsequent paper we will use the knowledge of the evolution of the distribution of lengths and velocities of the semilocal segments and present the results of detailed comparisons of our simulations with the VOS model for semilocal strings, leading to its quantitative calibration.

II. THE MODEL

The simplest semilocal string model Lagrangian reads [22,23]

$$S = \int d^4x \sqrt{-g} \left[[D\Phi]^2 - \frac{1}{4} F^2 - \lambda \left(\Phi^\dagger \Phi - \frac{\eta^2}{2} \right)^2 \right], \quad (1)$$

where Φ is a doublet of complex scalar fields $\Phi = (\phi_1, \phi_2)$, $D\Phi = (\partial_\mu - iqA_\mu)\Phi$, $F^2 = F_{\mu\nu}F^{\mu\nu}$ and $F_{\mu\nu} = (\partial_\mu A_\nu - \partial_\nu A_\mu)$ is the gauge field strength. The model is invariant under $SU(2)_{\text{global}} \times U(1)_{\text{local}}$ [23], and after symmetry breaking, the particle content is two Goldstone bosons, one scalar boson with mass $m_s = \sqrt{2}\lambda\eta$ and a vector boson with mass $m_v = q\eta$. After suitable rescalings:

$$\Phi \rightarrow \frac{\eta}{\sqrt{2}} \Phi, \quad x \rightarrow \frac{\sqrt{2}}{q\eta} x, \quad A_\mu \rightarrow \frac{\eta}{\sqrt{2}} A_\mu, \quad (2)$$

and defining

$$\beta = m_s^2/m_v^2 = 2\lambda/q^2, \quad (3)$$

the action can be rewritten in the numerically convenient form

$$S = \int d^4x \sqrt{-g} \left[[D\Phi]^2 - \frac{1}{4} F^2 - \frac{\beta}{2} (\Phi^\dagger \Phi - 1)^2 \right], \quad (4)$$

where now $D\Phi = (\partial_\mu - iA_\mu)\Phi$.

Since our aim is to characterize the dynamics of a network of semilocal strings in the early universe, we consider a spatially flat Friedmann-Robertson-Walker space-time with comoving coordinates:

$$g_{\mu\nu} = a(\tau)^2 \eta_{\mu\nu}, \quad (5)$$

where $\eta_{\mu\nu} = \text{diag}(-1, 1, 1, 1)$ is the Minkowski metric, τ is conformal time and $a(\tau)$ is the cosmic scale factor. The scale factors we will consider are those corresponding to a radiation dominated universe ($a \propto \tau$) and to a matter dominated universe ($a \propto \tau^2$).

The equations of motion for the semilocal model in the temporal gauge ($A_0 = 0$) read

$$\begin{aligned} \dot{\Pi} + 2\frac{\dot{a}}{a}\Pi - \mathbf{D}^2\Phi + a^2\beta(|\Phi|^2 - 1)\Phi &= 0, \\ \partial^\mu F_{\mu\nu} - ia^2(\Phi^\dagger D_\nu \Phi - D_\nu \Phi^\dagger \Phi) &= 0, \end{aligned} \quad (6)$$

where $\dot{} = d/d\tau$ and $\Pi = \dot{\Phi}$, together with

$$\partial_i F_{0i} = ia^2(\Pi^\dagger \Phi - \Pi \Phi^\dagger), \quad (7)$$

which is Gauss's law.

Semilocal strings can be thought of as Abelian Higgs strings embedded into a larger model. Indeed, if we set one of the scalar fields in the Φ doublet to zero, we recover the Abelian Higgs model exactly. The semilocal model can also be thought of as a limiting case of the electroweak model where the U(1) symmetry becomes global, and semilocal strings are thus related to the Z-strings of the electroweak model [37–40].

Unlike the Abelian Higgs strings, semilocal strings are not topological. Therefore, semilocal strings do not necessarily have to be closed or infinite, but they can form finite string segments: they can have ends. The semilocal string segments are linelike concentrations of magnetic flux, and the string ends can be understood as some sort of global monopoles, with long range interactions [33]. Note that these monopoles are not *true* global monopoles; they are, after all, string ends that behave *similarly* to global monopoles. Our analysis provides further supporting evidence for this interpretation.

This combination of string plus monopole provides semilocal strings with very rich dynamics: they can shrink and disappear, they can join neighboring segments to create larger ones, or the ends of a segment can join each other to form a loop. Besides their more complicated dynamics (compared to Abelian Higgs strings or monopoles), the fact that semilocal strings are nontopological has also important consequences for the numerical detection of their spatial location: there is no topological constraint that forces the

scalar field to be zero at the core of the string, and also there is no unique winding associated with the location of the string. One gauge invariant quantity that we will use to decide where a semilocal string lies is the concentration of magnetic energy.

The stability of semilocal strings is not trivial. Being nontopological, their stability is not warranted by topology but is instead dependent on dynamical and energetic considerations. It is controlled by the parameter β , c.f. Eq. (3). Semilocal strings are stable for $\beta < 1$, neutrally stable for $\beta = 1$ and unstable for $\beta > 1$ [41].

III. NUMERICAL SIMULATIONS

A. Numerical setup

Our main aim is to characterize the network of semilocal defects. In order to do so, we have obtained the equations of motion from the discretized version of the Hamiltonian corresponding to the action (4), and solved them using standard techniques (lattice-link variables and a staggered leapfrog method) in 1024^3 lattices with periodic boundary conditions, as explained in [34,36,40]. The simulations were performed at the COSMOS Consortium super-computer and i2Basque academic network computing infrastructure.

Since we have periodic boundary conditions, the simulation time has an upper bound, given by half the light-crossing time; for times longer than that, spurious boundary effects kick in. As is generally the case with this type of numerical simulation, there is a trade-off between dynamical range, spatial resolution and accuracy. One wishes to simulate the dynamics of the system for as long as possible, and therefore needs boxes with a physical size as large as possible. But one also wishes to simulate the equations as accurately as possible, which means that the lattice spacing has to be small. Thus, a compromise must be reached. We have chosen to use a lattice spacing of $\Delta x = 0.5$ and a time-step of $\Delta \tau = 0.2$, as our optimal balance between accuracy and dynamical range.

We wish to simulate semilocal string network evolution in the early universe, focusing on the radiation and matter domination eras. Defect network simulations in expanding backgrounds present us with another difficulty: the physical size of the defects is fixed throughout the simulation, but the size of the simulation box is growing with time. Equivalently, viewed in comoving coordinates, the box size is constant, but the comoving size of the defects shrinks. Again, this would point to the necessity to simulate the system in larger lattices, but this is not feasible. Instead, this difficulty is overcome by employing the Press-Ryden-Spergel (PRS) algorithm [42] whereby the defect cores are made to “artificially” grow comovingly during the simulation. This is achieved by promoting the parameters of the model into time varying variables [26], so that the (continuum) equations of motion (6) get modified to:

$$\begin{aligned} \dot{\Pi} + 2\frac{\dot{a}}{a}\Pi - \mathbf{D}^2\Phi + a^{2s}\beta(|\Phi|^2 - 1)\Phi &= 0, \\ \partial^\mu(a^{2(1-s)}F_{\mu\nu}) - ia^2(\Phi^\dagger D_\nu\Phi - D_\nu\Phi^\dagger\Phi) &= 0, \end{aligned} \quad (8)$$

where the parameter s gives the level of modeling in the PRS algorithm: $s = 1$ gives the true equation of motion, whereas $s = 0$ corresponds to defects whose comoving core size is constant. Several studies in the literature [26,43–45] show that the $s = 0$ case is a good approximation to the true $s = 1$ situation, with errors often smaller than statistical uncertainties. In the present work we performed all simulations using $s = 0$.

We are interested in estimating the velocity of the network, and more precisely the velocity of the segments and of the string–ends (monopoles). But for the purposes of calibrating the VOS model, it is particularly useful to do this once the system has reached scaling. As mentioned in the Introduction, scaling is a convenient property for the reliable study of defect networks, since we need to extrapolate from our limited numerical simulation into the much larger timescales relevant for defect networks evolving in the real universe. For scaling networks, the details of the initial conditions are not important as all information about the initial condition is lost once scaling has been reached. However, choosing a “good” set of initial conditions is important in practice, as it can help the network reach scaling as fast as possible, so that the system can evolve in the scaling regime for as long as possible within our finite dynamical range. The initial conditions we chose are the same as those in [36]: the gauge field, gauge field velocities and scalar field velocities are set to zero. The scalar fields are chosen to lie in the vacuum manifold, but have randomly chosen orientations [46]. Note that one extra benefit of using the PRS algorithm with $s = 0$ is that scaling is reached faster, which further increases our usable dynamical range.

We want to simulate strings that are stable and lead to long enough segments, so that we can get a representative statistical ensemble in our simulations. As mentioned earlier, semilocal strings are stable for values of $\beta < 1$, and our simulations were performed for $\beta = 0.04, 0.09, 0.15, 0.20, 0.25, 0.30, 0.35$. Higher values of β show a much more scarce network [40], and it would be much harder to get a numerous enough ensemble of semilocal strings. Lower values of β are also not optimal, since the scalar string cores start to be too thick for our simulation parameters.

For every value of β , and for both cosmologies (radiation and matter), we have performed 7 different simulations, in order to increase our statistics. It is numerically very expensive to analyze and output the simulation data at every time-step; instead, the information is output every 20 time steps once the network has approached scaling. The times chosen to output the data range from $\tau = 96$ until $\tau = 256$; for times earlier than $\tau = 96$ the system has not

settled into scaling well enough, and $\tau = 256$ is our limit due to the aforementioned boundary conditions. For a couple of cases the information has been output more frequently too, in order to check that the time spacing between outputs was adequate, and also to pinpoint some issues that we will discuss later. The output information can be used directly to estimate some of the quantities of interest, and also can be treated further to infer other useful quantities, as explained in the next section.

B. Algorithms to detect segments and obtain their velocities

In this subsection we will describe the quantities that need to be monitored in order to characterize semilocal string networks. We will introduce algorithms for estimating string length, string velocities and monopole velocities; some of these algorithms have already been used previously in the literature and some are introduced here for the first time and are specific to semilocal strings. Note that this is a general description; various caveats will be described in more detail in the following section.

As mentioned earlier, the fact that semilocal strings are not topological makes it much more difficult to numerically detect their location in a simulated lattice compared to the (much more studied) AH strings. In the AH case, the strings have a topological constraint, and one can follow them throughout the simulated lattice by tracking the winding of the scalar field across each plaquette [47]. The points with winding coincide with the zeros of the scalar field and therefore the windings happen at the same points at which the potential energy is concentrated. Furthermore, those are also the regions where the density of magnetic energy is highest. Thus, one can use any one of these features (high magnetic field density, high potential energy density, zeros of the scalar field, winding of the scalar field) to detect an AH string.

In the semilocal case, there are twice as many scalar fields, and just one gauge field. It is not clear then which field one has to choose to follow the windings, or whether these windings actually give the position of the string. Besides, the field does not have to be zero when there is a winding. Unlike in the AH case, *one* field can be winding around a site, but the *other* field may still fulfill the requirements to be in the vacuum of the potential energy. Therefore, we can have a field winding, but no concentration of potential or magnetic field.

In previous works [34,36,40,48], the criterion used to decide whether a point in the lattice belongs to a semilocal string was based on the concentration of magnetic energy (we will describe this method in more detail below). The procedure gives a collection of points that can be grouped into segments by proximity, i.e., we end up with a *volume* of points, which is subsequently used to estimate length. This length can be used to check, for example, the scaling of the network of strings. Moreover, the number of

segments is also a good estimator for the number of monopoles, i.e., segment ends, which is a further diagnostic for the scaling of the network. It has always been understood that this was just a first approach to obtain the length of strings, and we will show later that this estimator was seriously underestimating the string length.

Instead of a volume characterization of the strings, one would rather get a one-dimensional representation of strings, using, for example, the position of the core of the string. We will introduce a new estimator using field windings that will produce such a one-dimensional characterization of the string. This will be invaluable for the estimators of velocity we will describe later, and will enable us to determine individual segment velocities for semilocal strings for the first time.

We have also detected monopoles directly from the simulation for the first time. The number and velocity of monopoles are also important quantities for characterizing our system. For example, from the total number of monopoles \mathcal{N} an independent measurement of the number of segments can be inferred.

In what follows we will describe the algorithms used to obtain lengths of segments, number of segments, number of monopoles; velocities of segments and monopoles, network velocity; and quantities used to monitor scaling.

- (i) Estimation of segment lengths using the threshold of the magnetic field:

One criterion for deciding whether a lattice-point in the simulation is part of a semilocal string is the following: points where the magnetic field is higher than a given threshold are considered to belong to a semilocal string. In order to get a meaningful threshold, for a given β we consider the corresponding

Abelian Higgs string, and calculate the maximum of its magnetic field B_{\max} . We also calculate the radius of the Abelian Higgs string r_{AH} , defined as the radius at which the magnetic field $B_{\text{threshold}}$ (absolute value) drops to 30% of B_{\max} . Armed with the value B_{\max} , the whole simulation grid is scanned at every time-step, the value of the magnetic field is calculated at each grid point, and if the magnetic field is higher than that of $B_{\text{threshold}}$, the position of that point is output.

Once all the points have been output, they are distributed into segments. In order to do that, the points that are adjacent to each other are grouped into one segment. Thus, a collection of points forming the volume of a segment is obtained. Dividing that volume by the cross section area of the string (calculated from r_{AH} assuming a circular cross section), the length of the segment is estimated. From this procedure a distribution of segment lengths is obtained at every time step. Adding up all the segments, the total length of string in every time step $\mathcal{L}(\tau)$ can also be obtained, which can be used, for example, as a measure of scaling.

The left panel of Fig. 1 shows the positions of the segments in a typical simulation box obtained by plotting the points with a magnetic field higher than 30% of B_{\max} . This was the procedure used in previous papers to obtain semilocal segment lengths [34,36,40].

- (ii) Estimation of segment lengths using the windings of the scalar fields:

This is a new length estimator which allows us to obtain a one-dimensional characterization of the segments. In order to do so, it uses the windings in either or both of the scalar fields (ϕ_1 and ϕ_2) in the model.

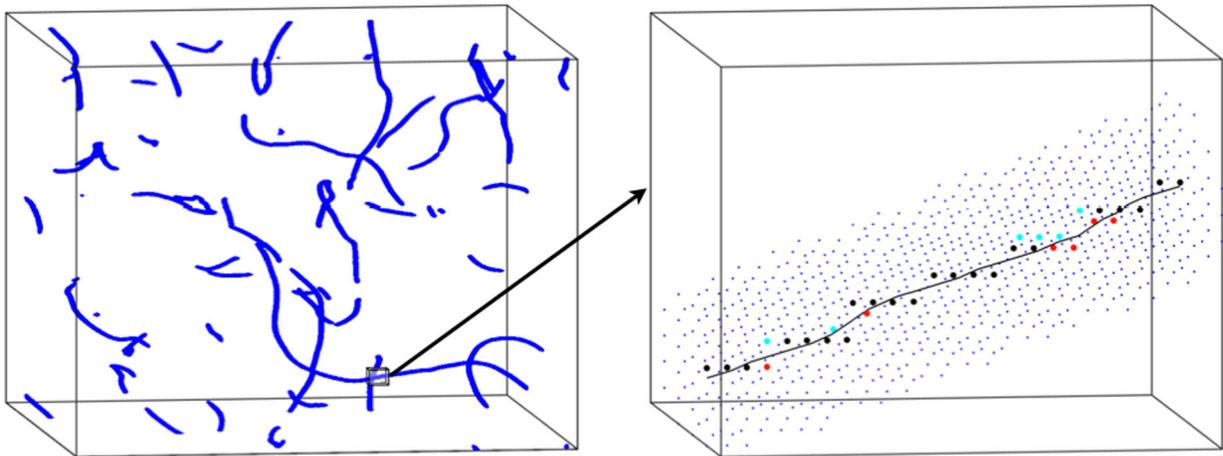


FIG. 1. In the left panel, the string segments in a typical simulation box are shown; the blue points are points with a magnetic field higher than the threshold described in the text. The right panel zooms in on a portion of one segment. The blue dots correspond again to points with high magnetic field, and the red, cyan and black points correspond to points where the windings of the scalar fields ϕ_1 , ϕ_2 or both (respectively) are nontrivial. Notice how these dots are not nicely aligned in a smooth curve, but jitter around (this is the Manhattan effect). Performing the smoothing over four nearest neighbors on each side of each point yields the black line, which is the smoothed version of the center of the core of the segment. Actually, the black line is also formed of points, but here those points have been joined by a line to show how they interpolate between the lattice points with winding.

The procedure is as follows: During the simulation, the winding of fields ϕ_1 and ϕ_2 is calculated at every point, and if this winding is different from zero, that position is output together with the value of the magnetic field at that point. The value of the magnetic field is needed because having the fields winding around some plaquette does not necessarily mean that point belongs to a string segment; it is only when the nonzero winding happens in a region with a high concentration of magnetic field that the position is regarded as the core of a string. The position of the core is given by the non-zero winding of either (or both) of the scalar fields (within the cloud of points with magnetic field), and following the points with winding one can map the 1-D array of points defining the centre of the string.

Actually, since there is no topological constraint and we are following the points with nonzero windings of one of the fields, the points may stop belonging to a semilocal string segment (because there may not be a high concentration of magnetic field), or the sequence of points with windings may abruptly stop. It is worth noting that, in principle, for a point to be part of a semilocal string it is enough that one of the fields winds in a region with high magnetic field. It is not necessary that the other field also winds. (Nevertheless, we will show later on that both fields seem to wind inside the string core.) On the other hand, sometimes both fields may be winding inside a cloud of magnetic field, but (due to discretization issues) not exactly on the same plaquette and they may be displaced by one lattice unit. In order to overcome these nuisances, the points of nonzero windings of both fields are combined. Those points that share neighbors are grouped, and that is how we define a segment. Moreover, those points can now be ordered. Every segment is thus formed of a (one-dimensional) ordered collection of points where either (or both) of the scalar fields have a nonzero winding and magnetic field is above some threshold, and points are in contact with each other by pairs. Figure 1 shows the string positions, obtained by taking into account the concentration of magnetic field (blue) and the points where the fields are winding (ϕ_1 in red, ϕ_2 in cyan and both in black).

Since our fields are discretized on a lattice, the determination of the center of the string suffers from the Manhattan effect [49]: the center of the segment is not a smooth curve, but it is actually formed by a collection of unit steps. A smooth version of the position of the cores would be beneficial as the estimation of both the length and the velocity of the segment would improve. Thus, the points in the string segments are smoothed by averaging the position over nearest neighbors. After trials using different number of neighbors, averaging over 4 neighbors at

each side seems optimal: the position of the core is smooth enough, and the structure of the segment is not lost (as would happen by averaging over too long distances).

The outcome of this procedure is a collection of segments, given by a one-dimensional list of the smoothed position of their core. The right panel of Fig. 1 depicts a zoom into a portion of a segment, showing how the averaged version of the core of the string, obtained by smoothing, nicely interpolates between the points with nonzero windings.

(iii) Estimation of the velocity of the segments:

Once the segments have been characterized by a smooth one-dimensional array of positions, this set of data can be used to track the history of each segment in time, and also estimate its (transverse) string velocity.

Consider a string segment at time τ_1 , and the same segment at a later time τ_2 . If one is able to estimate where each point in the segment has moved from τ_1 to τ_2 , we can get an ensemble average. This can be done in the following way: choose a point x_1 in the core of a segment at time τ_1 , and find its distance to all points belonging to the segment's core at τ_2 . The point at time τ_2 that is closest to x_1 is identified as the point where x_1 has moved to from τ_1 to τ_2 .² The velocity of point x_1 is then estimated by merely dividing the distance it has traveled by the time interval $\tau_2 - \tau_1$. Performing this operation with every point belonging to a given segment at every time step allows us to obtain a segment velocity (an average of $|v|$ over all the points in the segment). The average of all the segment velocities, which is not length-weighted, would then be the network average velocity.

Although this procedure has some subtleties, which we will describe in the next section, the benefits are manifold: one can obtain individual segment velocities at every time step. We can thus record the history of the velocity evolution of each segment in the simulation. Moreover, a by-product of this procedure is that one can create a map of which segments merge with which.

It is clear that for this procedure to work a one-dimensional characterization of the string segment is necessary, and therefore the velocity determination of segments relies on the string length characterization given by the windings.

(iv) Estimation of number and velocity of string-ends (monopoles)

As mentioned in Sec. II, the field configurations at the ends of the segments can be understood as global monopoles [23] and, as such, they can be detected in

²We do not remove points at the later stage, so that it may happen (and it will clearly happen for example in a collapsing loop) that two or more points at τ_1 may move to the same point τ_2 .

a simulation directly from the fields. The whole simulation lattice is scanned point by point to check whether monopoles are present. In order to do so, following [23], the field configurations at the string ends are recast from the 4 real scalar fields with SU(2) symmetry into three scalar fields, establishing an analogy to $O(3)$ global monopoles, through

$$\Psi \sim \Phi^\dagger \vec{\sigma} \Phi, \quad (9)$$

where $\vec{\sigma}$ are the Pauli matrices. Actually, the field configuration thus obtained is still quite noisy to clearly detect monopoles, and we clean it further via the transformation

$$\Psi \rightarrow \tilde{\Psi} = \Psi - \nabla \times (\nabla \times \Psi).$$

Once the field configuration has been treated, the location of a monopole is determined by calculating the topological charges given by the surface integral [3]

$$N = \frac{1}{8\pi} \oint dS^{ij} |\tilde{\Psi}|^{-3} \epsilon_{abc} \tilde{\Psi}^a \partial_i \tilde{\Psi}^b \partial_j \tilde{\Psi}^c, \quad (10)$$

where dS^{ij} is the infinitesimal surface element. Since we are in a discretized environment, Eq. (10) cannot be used directly, and instead a discretized version is used (see Appendix B in [50]). Note that N can be positive (monopole) or negative (antimonopole). It is not to be confused with \mathcal{N} , the total number of monopoles plus antimonopoles, i.e., the total number of segments' ends.

When scanning through the simulation box, if the topological charge is different from zero, the position of the monopole together with its topological charge is output. One can then directly obtain the number of monopoles (or string ends) in the simulation lattice, which is an independent way of measuring the number of string segments: the number of segments would be roughly $\mathcal{N}/2$, although not exactly so because some segments can be closed loops and thus have no ends. Moreover, the number of monopoles is another estimator for checking the scaling regime.

Since the position of the monopoles (and their topological charge) is known at every time-step, the velocity of monopoles can also be estimated following a procedure similar to the one described above for segment velocities [45]: A monopole M_1 is chosen at time τ_1 , and its distance with respect to all monopoles at the next time τ_2 is calculated. Then, M_1 is identified with the closest monopole at time τ_2 . Repeating this procedure at all times, the history of M_1 can be tracked and an estimate of its

velocity can be obtained by merely dividing the distance it has travelled by the corresponding time interval.

Monopole velocity estimations also have their issues, but once again the benefits are manifold: one can obtain individual monopole velocities, both at each time step and as the average velocity during the life of the monopole. By averaging over all monopoles, a monopole-network velocity can be obtained too.

- (v) Estimation of the network velocity using local lattice variables.

There are also other types of velocity estimators for the whole network, based on local field quantities as described in [51–53]. These estimators are computed by considering a string at rest, and performing a Lorentz boost to it. Thanks to the different properties of each term in the Lagrangian under boosts, two different estimators can be written down:

$$\begin{aligned} \langle v^2 \rangle_F &= \frac{\mathbf{E}_{\mathcal{W}}^2}{\mathbf{B}_{\mathcal{W}}^2}, \\ \langle v^2 \rangle_G &= \frac{2G_{\mathcal{W}}}{1 + G_{\mathcal{W}}}, \end{aligned} \quad (11)$$

where F refers to the gauge field strength $F_{\mu\nu}$, and

$$G_{\mathcal{W}} = \frac{\Pi_{\mathcal{W}}^2}{(\mathbf{D}\Phi)_{\mathcal{W}}^2}. \quad (12)$$

The subscript \mathcal{W} denotes weighting by some appropriate physical quantity. In the present case, a magnetic energy weighting was used, because this automatically ensures that only regions with non-vanishing magnetic energy contribute to velocities. Regions with semilocal strings have higher concentration of magnetic energy, and thus those regions contribute most to the above integrals. For a given quantity A the weighting is applied in the following way:

$$A_{\mathcal{W}} = \frac{\int d^3x A \mathcal{W}}{\int d^3x \mathcal{W}} \quad (13)$$

Using this definition for a *weighted average*, the estimations above read

$$G_{\mathcal{W}} = \frac{\Pi_{\mathcal{W}}^2}{(\mathbf{D}\Phi)_{\mathcal{W}}^2} = \frac{\int d^3x \Pi^2 \mathcal{W}}{\int d^3x (\mathbf{D}\Phi)^2 \mathcal{W}}$$

for example. During the simulation, the quantities needed to obtain the velocity estimators in (11) are output.

Note that these field velocity estimators provide information on the network as a whole, whereas the

previously described methods for segments (and monopoles) give information about each individual segment (or monopole). Besides, the estimators in Eq. (11) average over all regions with some magnetic energy, and thus both segments and monopoles contribute, as well as regions where a segment may have disappeared (or it is about to disappear) and a temporary magnetic field density is left-over.

(vi) Monitoring scaling:

Scaling is a key property of a network of defects, which is indispensable if one needs to extrapolate the results of a numerical simulation (whose dynamic range will necessarily be small for the scales of interest) into the history of a realistic defect network. Scaling can be measured by the total string length $\mathcal{L}(\tau)$ and by the number of monopoles $\mathcal{N}(\tau)$ (which is related to the number of segments, $\approx \mathcal{N}(\tau)/2$). Besides, the energy (T_{00}) of the system is also a good candidate to monitor the scaling of the network.

IV. CAVEATS AND DIFFICULTIES OF THE ALGORITHMS AND THE NUMERICAL SETUP

In this section we will describe some caveats and difficulties we encountered in the simulations. These, on the one hand, have to be dealt with in order to obtain physically robust results, and on the other, provide insights into how to further extend the analytic model for semilocal strings (the last point will be addressed in a subsequent paper).

A. Comparison between the two length estimators

In the previous sections, two different procedures for obtaining the length of string segments have been described: one using the threshold of the magnetic field and the other using the windings of the scalar fields. We did expect some mismatch between these two procedures, mostly because the method of obtaining segment lengths by counting points above a magnetic field threshold and dividing the *volume* by the *cross-section area* was a rather crude algorithm. This algorithm was used in previous works [34,36,40] due to the difficulty of defining a semilocal string otherwise (with windings or zeros of the scalar field), and was believed to be a good first approximation.

Figure 2 shows the relation between the lengths of the segments measured by using the magnetic flux and the winding approach for $\beta = 0.20$ at $\tau = 256$. The ratio between these two quantities is almost constant, and it is roughly a (surprisingly large) factor of 1.5 for most values of β both in the matter and radiation cases. The factor is a bit different for shorter segments, which is presumably due to the fact that the segment ends play a stronger role for shorter segments, but this is not relevant for the whole network length, which is dominated by long segments.

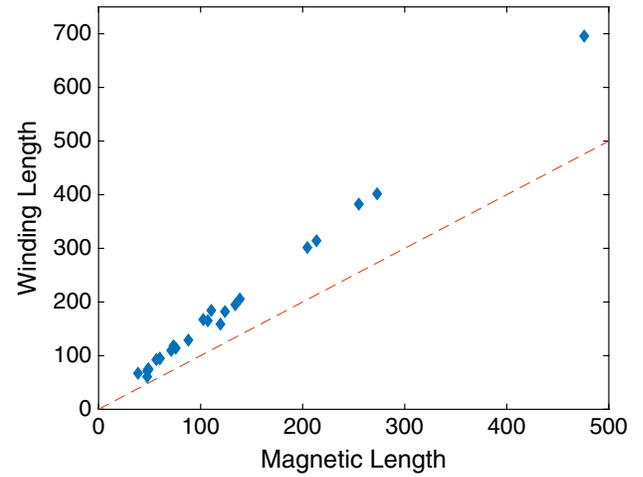


FIG. 2. In this figure the relation between the two length-measuring approaches is shown. The horizontal axis has the lengths measured with the magnetic flux approach, while the vertical axis has the lengths measured with the winding approach. Every diamond represents a segment in a simulation box for $\beta = 0.20$ at time $\tau = 256$. It is clear that the proportionality is almost the same for all the segments in the network. The diagonal is shown with a dashed line, emphasizing the bias factor between the two approaches.

There are several possible sources of uncertainty that can help us understand the difference in the string length obtained with the different procedures. First of all, previous determinations discarded small “blobs,” very short segments whose length is below some threshold (a few times the core radius). Second, the fact that cross sections are estimated from the Abelian Higgs profile could lead to an overall bias. There are also two possible physical effects to take into account: one is Lorentz contraction of the segments due to their velocities, making segments narrower than the estimated cross-section of the string; the other is that segments are not straight, and when they bend the strings can be considerably narrower.

There is also a clear numerical uncertainty in the procedure of obtaining the cross section of the string. Recall that the cross sections of the semilocal strings in the simulations are defined to be the same as those of a straight AH string at rest. The typical radius for AH strings range from $r_{04} = 2.36$ for $\beta = 0.04$ to $r_{35} = 1.71$ for $\beta = 0.35$. The lattice spacing we are using is $\Delta x = 0.5$ which already gives us a considerable error on the string radius: we consider that points with magnetic field higher than 30% of the maximum magnetic field belong to the string. Imagine a string centered at a lattice-point, with radius, say, 1.4. The string radius then does not quite cover 3 lattice units ($3 * \Delta x > 1.5$), and therefore there are only 2 lattice points which *qualify* as points of strings, but the volume of strings obtained like that will nevertheless be divided by the *true* radius, clearly underestimating the string length. Moreover, a string will typically not be centered around a lattice-point,

but it will be located anywhere within a lattice-cell, making this effect more important. Thus, the points that we *numerically* decide that belong to the string are *fewer* than what a more sophisticated procedure would get. We can get an estimation of the maximum errors for the two extreme β as follows:

$$\frac{r_{04}^2}{(r_{04} - \Delta x)^2} = 1.60, \quad \frac{r_{35}^2}{(r_{35} - \Delta x)^2} = 1.99.$$

This numerical bias seems to account for most of the discrepancy between the two length estimation procedures, while the Lorentz factor and the fact that strings are not straight do not seem to be so important. Actually, the Lorentz contraction of the segments is clearly not a significant factor to take into account since the velocity values we obtain later on are rather low to obtain a Lorentz factor capable of explaining the difference. In Table I, the total length in segments for every β is shown. We conclude then that the string segments length estimated in previous works by (some of) us has been overestimated by roughly a factor of 1.5 (the average ratio between the values in Table I are 1.56 for Radiation and 1.55 for matter).

When using the approach based on obtaining volumes of points and then dividing by string cross sections, there was also some nuisance with very small segments, which were dubbed *blobs* in [36]. These were very small regions with some magnetic flux, possibly as a leftover of a recently collapsed segment. There was some uncertainty in how to remove those blobs. With the new winding approach, this problem of blobs has disappeared. In any case, we have checked that the number of monopoles obtained by the winding approach, and by the magnetic field approach (removing blobs) is very similar.

One very nice check that our new length estimators work well can be found in Fig. 3. There, we plot the positions of the strings obtained by the winding procedure, and the positions of the monopoles obtained by calculating the

TABLE I. The total lengths in $\tau = 256$ for every β we simulated and for radiation and matter epochs. L_{winding} denotes the length estimator using the windings of the scalar fields, whereas L_{magnetic} denotes the length estimator using a threshold of magnetic field. Note that these numbers correspond to a single simulation per value of β .

β	Radiation		Matter	
	L_{winding}	L_{magnetic}	L_{winding}	L_{magnetic}
0.04	6166.44	3895.33	12303.07	8014.28
0.09	6903.84	4673.61	10780.82	7266.59
0.15	5246.97	3476.99	9267.51	6099.45
0.20	5221.27	3546.95	7447.93	4865.46
0.25	4450.95	2948.52	6147.42	3853.46
0.30	2546.31	1585.21	4562.78	2901.85
0.35	2095.5	1173.86	3405.10	2119.90

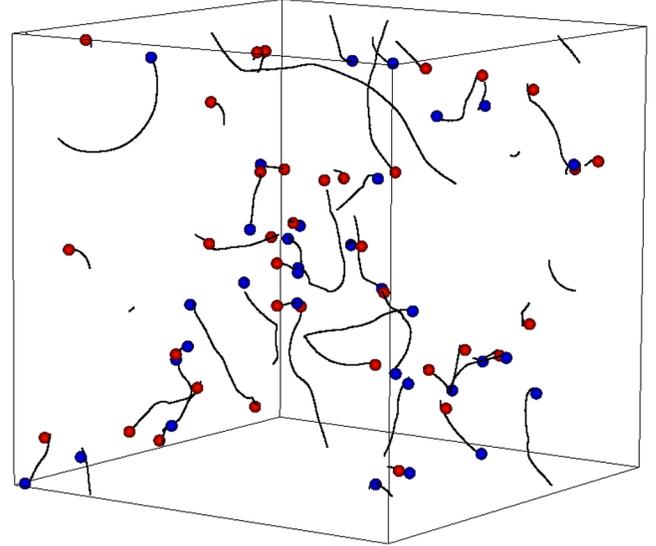


FIG. 3. In this figure one can see the box after the characterization. The black lines represent the string segments detected using the windings, where each point is smoothed using the four nearest neighbors in each direction. The red dots represent the antimonopoles and the blue ones the monopoles.

monopole charge. In principle, these two procedures are independent, but the positions of the strings and monopoles (string-ends) are aligned very nicely.

From the length estimators (of both monopoles and strings) two different VOS-type length scale parameters can be obtained:

$$\gamma_{\mathcal{L}} \equiv \frac{1}{\tau} \sqrt{\frac{V}{\mathcal{L}}}, \quad \gamma_{\mathcal{M}} \equiv \frac{1}{\tau} \left(\frac{V}{\mathcal{N}} \right)^{1/3} \quad (14)$$

where \mathcal{L} is the estimation of the string length, and \mathcal{N} is the estimation of number of monopoles (double the number of segments). We show the values of those parameters in Table II in Sec. V, but we would like to point out that these values are compatible (within errors) with the values obtained in [36], for both $\gamma_{\mathcal{L}}$ and $\gamma_{\mathcal{M}}$. Note that, in view

TABLE II. Values of the VOS-type length estimators for the total length of segments $\gamma_{\mathcal{L}}$ and the total number of monopoles $\gamma_{\mathcal{M}}$ (14) in the box for every β .

β	Radiation		Matter	
	$\gamma_{\mathcal{L}}$	$\gamma_{\mathcal{M}}$	$\gamma_{\mathcal{L}}$	$\gamma_{\mathcal{M}}$
0.04	0.38 ± 0.05	0.48 ± 0.04	0.28 ± 0.02	0.41 ± 0.02
0.09	0.40 ± 0.04	0.43 ± 0.02	0.33 ± 0.02	0.38 ± 0.02
0.15	0.47 ± 0.09	0.40 ± 0.02	0.40 ± 0.04	0.35 ± 0.03
0.20	0.52 ± 0.08	0.41 ± 0.04	0.46 ± 0.04	0.37 ± 0.02
0.25	0.61 ± 0.09	0.42 ± 0.05	0.54 ± 0.06	0.37 ± 0.02
0.30	0.81 ± 0.09	0.46 ± 0.05	0.65 ± 0.09	0.37 ± 0.03
0.35	1.06 ± 0.09	0.49 ± 0.05	0.78 ± 0.09	0.41 ± 0.03

of our results, the values for $\gamma_{\mathcal{L}}$ in [36] should be corrected by a $\frac{1}{\sqrt{1.5}}$ coming from the factor 1.5 difference in the length estimator, but even without the factor, the values lie within $1-\sigma$ from each other.

In the rest of this work, we will only use the length estimator which relies on the windings approach. The reason for calculating segment lengths using the threshold of the magnetic field was for ease of comparing with previous works; it will not be used in the rest of this paper.

B. Identification of semilocal segments

As has been already explained, the position of a semilocal string is difficult to pinpoint because there is no topological obstruction for the scalar to acquire a nonzero value around a winding (one field may wind and the other may climb the potential) and there is no requirement for semilocal strings to be closed or infinite. Actually, it is *finite semilocal string segments* that we are studying.

One way of detecting the strings, explained above, is by following the windings of both scalar fields ϕ_1 and ϕ_2 , and checking where those windings happen within regions of high magnetic field. In the left panel of Fig. 4 we show the windings of both fields in the simulation box, without taking into account the values of the magnetic field: windings of ϕ_1 in blue and of ϕ_2 in green. Actually, there are several regions where the two fields wind in the same plaquette (or at a difference of 1 plaquette), and those are plotted in red.

The first thing we notice is that each of the fields (ϕ_1 and ϕ_2) creates a network of closed or infinite “strings,” very much like an Abelian Higgs network. But unlike in the Abelian Higgs case, a winding of a scalar field does not necessarily mean that there is a string forming; there is not necessarily a corresponding concentration of magnetic or potential energy. Where do semilocal strings form then? In

the right panel we plot the same windings, but also the regions where there is a high concentration of magnetic field. This shows that semilocal strings are actually the regions where *both* fields wind, which makes sense because in that case the scalar field has to be approximately zero there and therefore the magnetic field can be large in those regions.

Note that, although this network formed from points with windings of ϕ_1 , points with windings of ϕ_2 and semilocal strings is reminiscent of a p-q string network [21,54], where the Y-junctions correspond to monopoles, this analogy is misleading. The winding of a single field does not necessarily imply any physical concentration of energy, so there is no string tension that can be associated unambiguously with these “strings.”

C. Apparent superluminal velocities

The preliminary results on velocities (for both segments and monopoles) obtained by most of the direct algorithms, as well as from estimators based on local field quantities, cf. Eq. (11) give rather modest velocities. However, on a few occasions, we obtained values higher than the speed of light c ; in some very few extreme cases as large as $10c$. In principle, as we explain below, this need not signal a breakdown of causality, and it can be an artefact of the way we measure velocities. The dynamics is local and causal at the level of the fields.

A more detailed analysis showed that superluminal values for velocities were only obtained when using the estimators (described above) which try to track the segments and the monopoles in each time step as the network evolves. We also found out that the procedure needed some more careful handling, due to the following caveats:

(i) Segment velocities:

Imagine a segment that has closed into a loop at time τ_1 , and disappears before the next time τ_2 when

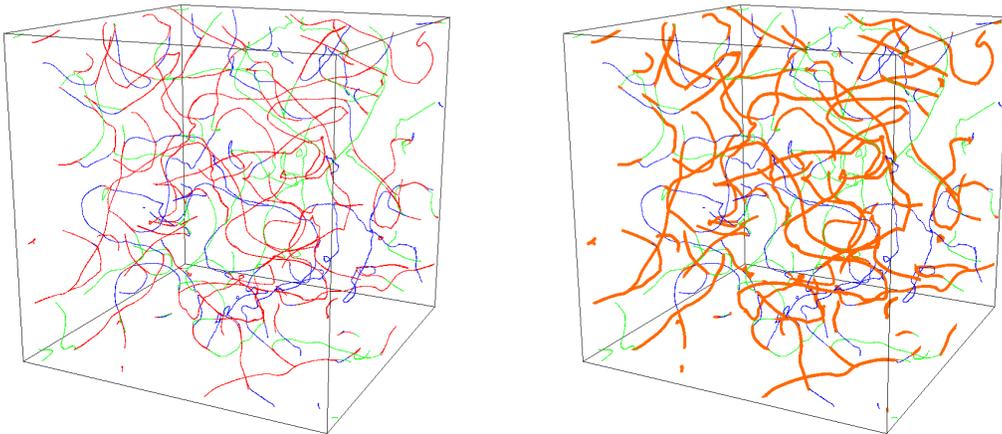


FIG. 4. In the left panel the windings of both fields are shown, without taking into account the magnetic field: blue for windings of ϕ_1 , green for windings in ϕ_2 and red for windings in both. As one can see each one of the fields creates a network of closed or infinite strings. On the right panel, we plot the same windings, but also the regions where there is concentration of magnetic field (orange). This figure shows that semilocal strings are actually the regions where both fields wind.

the box will be reanalyzed. Then, all the points belonging to the loop at time τ_1 will not have a close-by point to match with at τ_2 . However, our code will find the closest segment at time τ_2 , even though that will correspond to another segment, presumably further than causality allows. Therefore, the calculated velocity for the loop will be very high, but clearly it is not physical; it is a shortcoming of the procedure. One way of dealing with this is by applying a cutoff, and disregarding all velocities higher than that cutoff, but this has the danger of possibly losing some dynamics (if the cutoff is too strict), or considering too many unphysical cases (if the cutoff is too weak). Another possibility to avoid these problems is by performing the identification between segments at different times backwards, i.e., choosing a point at τ_2 and looking at all points at τ_1 to match it to.

However, if we revert to the *backward* identification, we may run into another problem: Imagine now a segment that is going to join to another segment, and for the sake of explaining this issue assume that, with the exception of the segment's ends (which are traveling towards each other), the string is at rest. At time τ_1 the two segments are separated by some distance, but at time τ_2 all that distance is filled with string. All the points that have filled that gap at time τ_2 will have their closest point at the ends of the segments at time τ_1 and therefore the velocity we get for those points is not the velocity of the segment. If at all, that velocity will be related to the segment end (monopole) velocity. This is avoided by calculating velocities *forward* (from τ_1 to τ_2).

Clearly, choosing the forward or backward approach may solve one of the two previous problems, but not both.

(ii) Monopole velocities:

Throughout the simulation monopole-antimonopole pairs annihilate: if a monopole M_1 annihilates with an antimonopole between times τ_1 and τ_2 , our procedure will not have the information to know that monopole M_1 has annihilated. Instead, the procedure will try to match it to another monopole (the one nearest to it), and the velocity obtained will be incorrect (and possibly supeluminal).

Other types of misidentification issues may also occur occasionally: There are some detection problems when a string segment collapses into itself, as the fields are reconfiguring to the new situation and radiating energy away. In this case, the field configuration at the string ends is related to the merging of a monopole and an antimonopole, and our detection algorithm fails (either by detecting spurious monopoles, or by failing to detect any charge at all).

Moreover, if at the moment when the identification algorithm is run a monopole passes through a face of the lattice, the algorithm may miss to detect it, because the topological charge will be divided into two different cells of the lattice. The way we have chosen to overcome this difficulty is by checking whether the monopole can be matched to a monopole in a subsequent time-step τ_3 , and to compare the distances obtained. If the distance to the monopole at τ_3 is smaller than to that at τ_2 (and is actually physical), the intermediate step is bypassed. The fact that the charge of the monopole is recorded is of help when there are monopoles and antimonopoles nearby: the topological charge of a monopole does not change during its evolution, i.e. a monopole does not become an antimonopole, and therefore the number of possible candidates to match to a given monopole is halved.

After this refinement of segment and monopole velocity estimators, most of the instances where superluminal velocities were obtained have been cured. However, there remains some cases where superluminal velocities appear, corresponding to a few instances of merging of segments as mentioned above. These superluminal velocities appear at a few points of a string segment close to the end of the string, and also obviously at the monopoles. For the segments of strings, the superluminal velocities appeared only in the backward direction, and their effect is to increase artificially the velocity of the segment (since the segment velocity is an average of the velocities of the segment points).

We individually looked at those cases, in order to determine whether those high velocities were physical or had their origin on the shortcomings of our algorithms, and found that the reason for these apparently superluminal velocities was that the field configuration between the two monopoles that are about to merge is such that it is energetically favourable for a new segment to form in between the two advancing ones. In some sense, a new segment appears “out of thin air,” such that it seems that the monopole has instantaneously moved considerably. The monopole does not pull from the string and creates string as it moves; instead, a *chunk* of string is formed and the monopole makes a jump forward. One way to understand this is to think about the field dynamics in the plane orthogonal to the string that is about to form (see [55]). Consider the extreme case of a z -independent field configuration that is going to become a straight string in the z axis. Initially the magnetic field in the x - y plane is dispersed over a relatively large area, so its value is below the threshold where our algorithm would recognize the presence of a string, and it is accreting into a smaller area around the z axis—a region with a lower value of the scalar field than the surroundings. At some point the magnetic field grows above the threshold and we see the whole string appear in a single time step.

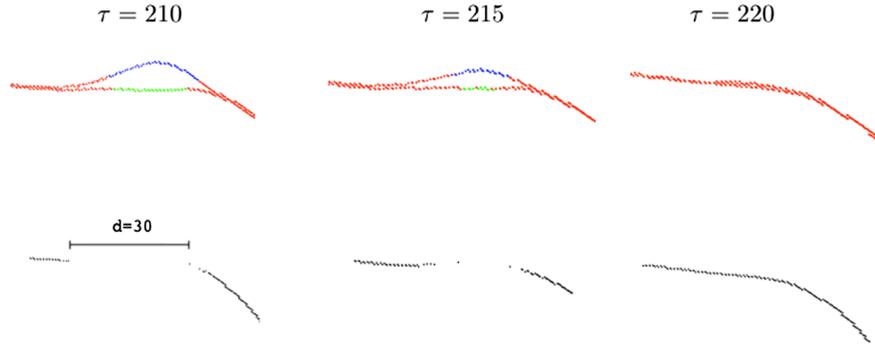


FIG. 5. In this figure we show the merging of two string segments, in what our algorithm would report as ultrarelativistic velocities. Initially the separation between both string-ends is $d = 30$ and in less than $\Delta\tau = 10$ both segments have merged. This process can be understood as small string segments being formed both close to the string ends, as well as in between both strings (depicted by those single points in between strings).

In other words, if the points of windings of ϕ_1 and ϕ_2 are almost parallel, it may be favorable for a collection of points to become part of a string all together. Since our simulation only takes snapshots every 20 time steps, this process can remain unnoticed, and the end result is that the segment has apparently grown superluminally, whereas the reality is that a new segment has been formed. This can also be understood by considering that the newly formed string segment has its corresponding monopole-antimonopole pair on its ends. The newly formed monopole (say) annihilates with the advancing antimonopole, and the newly formed segment merges with one of the old ones.

It is worth reminding the reader at this stage that we are *not* simulating segments and monopoles. We are simulating fields, and the monopoles and segments are consequences of them. Thus, a monopole does not really *move*, but it is the movement of the fields which leads to the apparent movement of the monopole. When a new segment forms, the physics of the fields is causal, but the consequence may seem to be a monopole moving ultrarelativistically, when, in essence, it is a different monopole.

Figure 5 shows one such situation. The two depicted fragments belong to segments that were behaving normally (i.e., not ultrarelativistically), until they got ready to merge. When they get close to each other, the dynamics of the fields leads to the merging as soon as possible: segment ends grow by creating small segments in the ends, and even in between the segments a new string is formed (shown as single points in the figure). Our algorithm is not able to catch these instances, and instead, velocities faster than light are reported.

The question remains now how to automatically factor out those instances, which would clearly corrupt the estimator and give much higher velocities than the true physical one. Learning from our previous experience on segment velocities, we apply the following method to disregard those unphysical velocities: For points that move very fast (in practice we chose velocities faster than $0.5c$), the velocities obtained by going “forward” (from τ_1 to τ_2)

and “backward” (from τ_2 to τ_1) are compared. If the difference is higher than 50% of the value of either of the velocities, we disregard it. We have tested that this is preferable to setting a hard cutoff, because the results clearly do depend on the cutoffs, and we do not want to artificially enforce that the velocities be subluminal in case this should result from the physics. The monopoles associated to these segment ends are also removed from the analysis.

V. NUMERICAL RESULTS

In this section we report on the results obtained from our simulations, after carefully accounting for the caveats described in the previous section.

We first checked that the simulations reached the scaling regime. Scaling can be checked by the evolution of the energy of the system, as mentioned earlier, by checking that $T_{00}^{-1/2}$ is linear with time τ . This is a measure that depends only on the values of the fields as simulated, without further treatment. We show the scaling on our simulations using the energy as a measure of scaling in Fig. 6.

There are also two other diagnostics for scaling that depend on the total length of string \mathcal{L} and the total number of monopoles \mathcal{N} . One can obtain a VOS-type length scale by the following combinations $\sqrt{\frac{V}{\mathcal{L}}}$ and on $(\frac{V}{\mathcal{N}})^{1/3}$. These quantities are derived, meaning that one has to extract this information from the simulation using estimators, and therefore, in some sense, they are more indirect than using the energy to check for scaling. We show the behaviour of these quantities in Fig. 7 for two extreme cases (radiation and $\beta = 0.04$) and (matter and $\beta = 0.35$). These curves show that the system does indeed reach scaling fairly early. The slope of the curves related to \mathcal{L} and \mathcal{N} is what was used above Eq. (14) to determine the VOS-type scaling parameters. The values of $\gamma_{\mathcal{L}}$ and $\gamma_{\mathcal{N}}$ can be found in Table II for every β studied.

One of the main results in this work is the values of the velocities of the semilocal network, both for strings and for

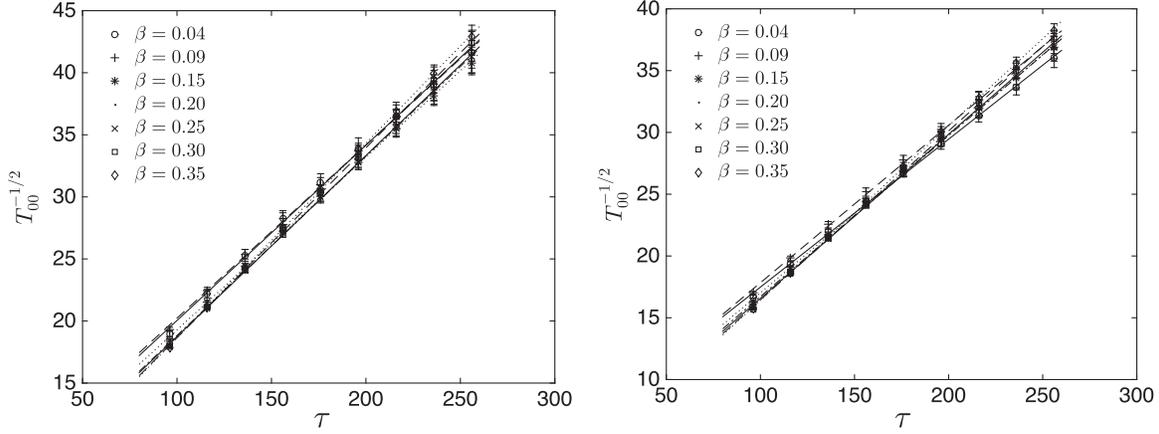


FIG. 6. This figure shows that our simulations reach a scaling regime for all β . The left panel shows the case of radiation domination, the right panel the case of matter domination.

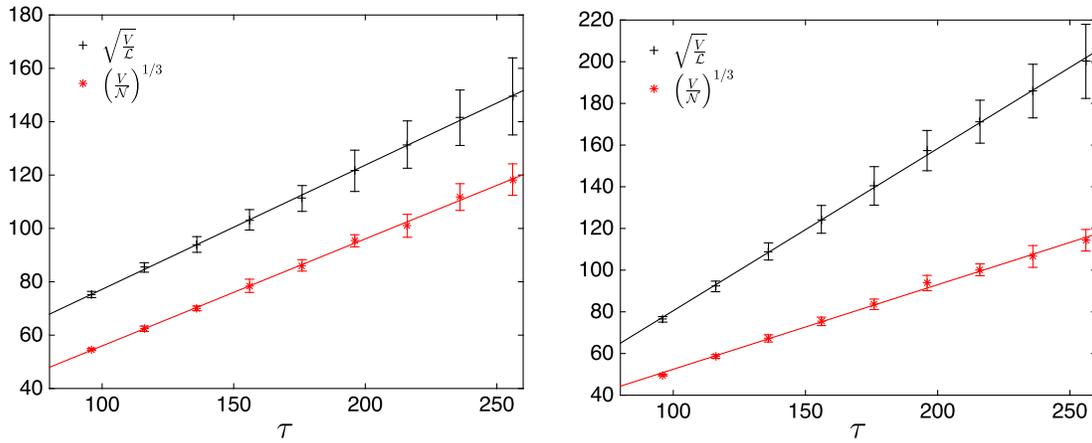


FIG. 7. Scalings in total string length and monopole number for $\beta = 0.15$ in radiation era (left) and for $\beta = 0.35$ in matter era (right).

monopoles, using our different estimators. In Table III we show the velocity values for every β in the radiation and matter dominated epochs obtained by following the positions of each string segment and monopole, and averaging

over all cases (the errors given are statistical errors). Overall one can see that the velocities in radiation are somewhat higher than the velocities in matter, as expected due to a lower damping term in the former epoch. More

TABLE III. Average velocities for segments ($v_{\mathcal{L}}$) and monopoles ($v_{\mathcal{M}}$) obtained by following the positions of the segments and the monopoles during the simulation. The errors are statistical errors obtained by averaging. The velocities in radiation are noticeably higher than in matter, but there does not seem to be a strong dependence on β for the velocities.

β	Radiation		Matter	
	$v_{\mathcal{L}}$	$v_{\mathcal{M}}$	$v_{\mathcal{L}}$	$v_{\mathcal{M}}$
0.04	0.345 ± 0.010	0.574 ± 0.010	0.266 ± 0.010	0.505 ± 0.010
0.09	0.338 ± 0.010	0.583 ± 0.010	0.265 ± 0.010	0.510 ± 0.010
0.15	0.337 ± 0.010	0.600 ± 0.012	0.262 ± 0.011	0.509 ± 0.010
0.20	0.337 ± 0.010	0.591 ± 0.010	0.260 ± 0.010	0.519 ± 0.010
0.25	0.337 ± 0.010	0.591 ± 0.010	0.261 ± 0.010	0.520 ± 0.010
0.30	0.342 ± 0.014	0.596 ± 0.010	0.259 ± 0.010	0.524 ± 0.010
0.35	0.337 ± 0.013	0.600 ± 0.012	0.262 ± 0.011	0.521 ± 0.012

TABLE IV. Velocities obtained from local field estimators, Eq. (11), for all β and radiation and matter dominations. Velocities in radiation domination are higher than in matter domination. Now there is moderately robust statistical evidence that the velocity of the network decreases for increasing β .

β	Radiation		Matter	
	v_F	v_G	v_F	v_G
0.04	0.57 ± 0.02	0.55 ± 0.02	0.41 ± 0.01	0.41 ± 0.01
0.09	0.57 ± 0.02	0.56 ± 0.02	0.42 ± 0.01	0.41 ± 0.01
0.15	0.55 ± 0.02	0.55 ± 0.01	0.41 ± 0.01	0.41 ± 0.01
0.20	0.53 ± 0.02	0.53 ± 0.02	0.40 ± 0.01	0.40 ± 0.01
0.25	0.51 ± 0.02	0.51 ± 0.02	0.39 ± 0.01	0.39 ± 0.01
0.30	0.49 ± 0.02	0.50 ± 0.02	0.38 ± 0.01	0.38 ± 0.01
0.35	0.47 ± 0.02	0.48 ± 0.02	0.36 ± 0.02	0.37 ± 0.02

interestingly, there does not seem to be a strong dependency of the velocities on β ; indeed, note that all these values are equivalent within their statistical errors.

In Table IV we show the velocities obtained by local field estimators, cf. Eq. (11). Note that these estimators do not distinguish between strings and monopoles, and give one

number for the average network velocity for semilocal strings as a whole. Once again, we obtain that velocities during radiation are faster than in matter; and we also note that both local field estimators give equivalent velocities. In this case there seems to be a trend, and velocities seem to decrease for increasing β .

The procedure of following the positions of the string segments and of the monopoles does not only yield a network velocity, but it also allows us to obtain the velocity of each segment and monopole. We can thus follow the history of each segment and monitor whether they merge with other segments, disappear, or just continue to evolve in the network. This is very interesting information from the point of view of the VOS model, because we can now tell if there is a correlation between string segment size and its velocity, for example.

In the subsequent figures we only report the results for $\beta = 0.04$ and radiation domination. The situation with other β and matter domination is analogous; figures for other values of β are provided as Supplemental Material [56] to this paper. The choice of $\beta = 0.04$ to show the results in the main body of the paper was made since the

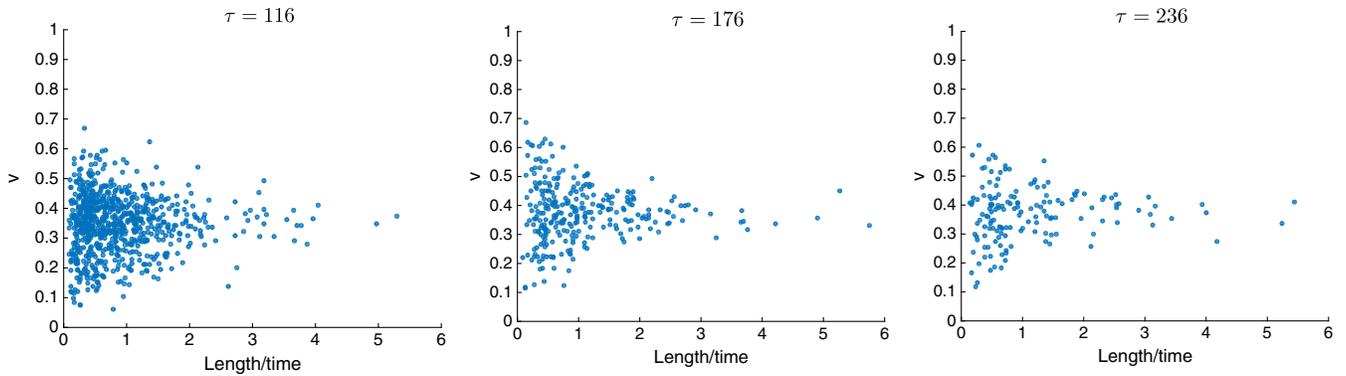


FIG. 8. This plot shows the distribution of segments with respect to velocity for simulation in radiation domination and $\beta = 0.04$ for all seven simulations. Each point represents a segment in the network, where in the x-axis the length of the segment divided by time is shown and its velocity in the y-axis. This is the case where segments *flow* through the network, i.e., they do interact with any other segment in the next time step.

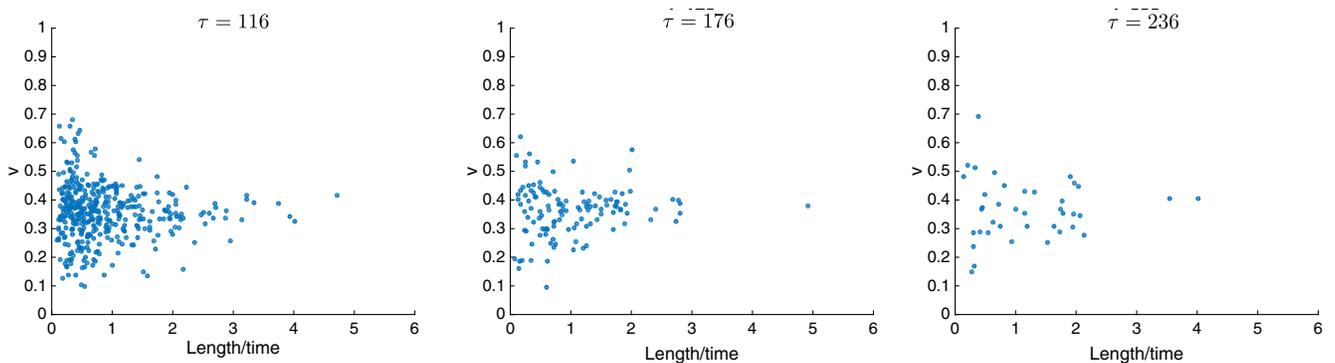


FIG. 9. This figure is similar to Fig. 8, but in this case the segments that are plotted are those that *merge* with other segments before the next time step.

number of segments for this β is higher, and thus the different phenomena can be more easily appreciated.

We show in Figs. 8 and 9 the velocity versus length distribution for some time steps. In these figures, each point

corresponds to a segment, and these are distributed according to their length divided by time (x axis) and velocity (y axis) for three different times ($\tau = 116, 176, 236$). These segments have been divided into two groups according to

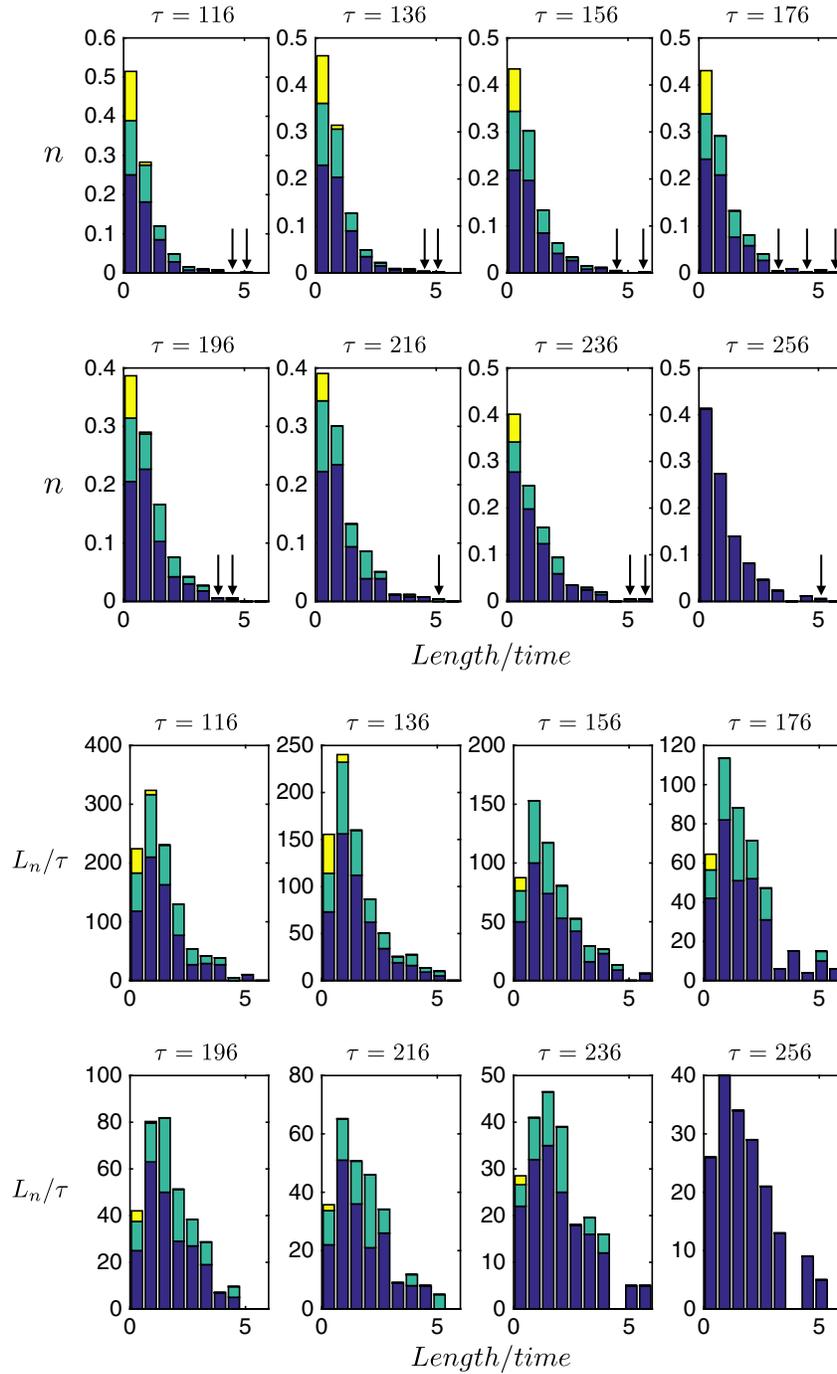


FIG. 10. These histograms show the distribution of the segments during the network evolution for $\beta = 0.04$ in the radiation dominated era for all seven simulations, where the segments are binned in 10 bins with uniform width. The top panels show the fractional distribution of the number of segments n with their lengths divided by time, whereas the bottom panels show the analogous distribution for the total lengths divided by time L_n/τ . The colors represent different types of segments, depending on their future behaviour: in blue segments which are *flowing*, in green segments that are *merging* and in yellow segments that are *collapsing* before the next time-step. We write arrows to remark that in those instances, there are a few (one or two) segments in that bin, which are hard to see in the top panels, but can be seen in the bottom ones. Note that in the last time step we have no information whether the segments will flow, merge or collapse, so we just choose to show them as flow segments.

their behavior *in the next time step*: On the one hand, in Fig. 8 we plot the segments which do not interact with other string segments in the next time step, i.e., segments which *flow* through the network. On the other hand, in Fig. 9 we

show the segments that do merge with other segments before the next time step.

It is worth mentioning that the points are scattered around a central value, which is roughly the same for all

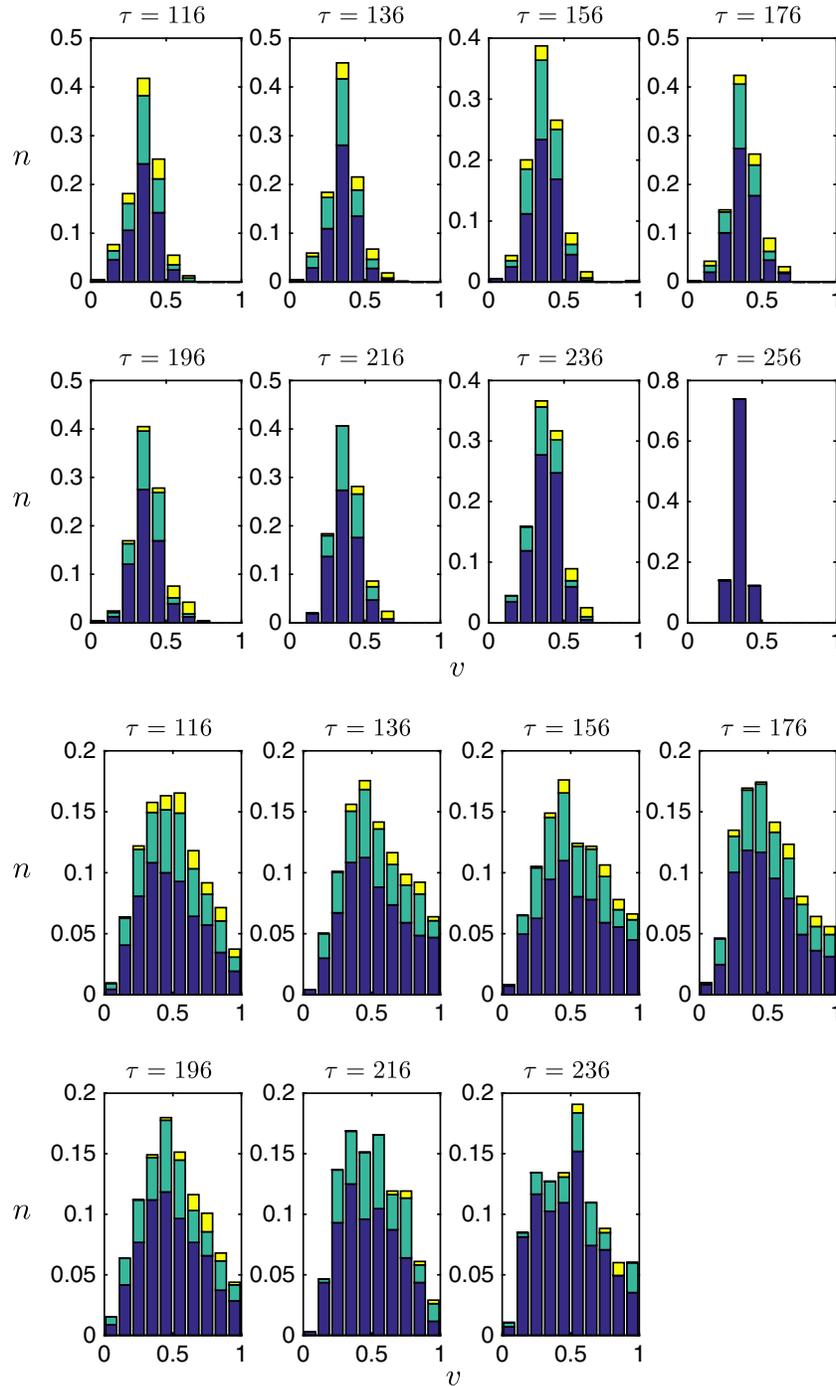


FIG. 11. These histograms show the velocity v distribution of the segments (top) and monopoles (bottom) during their evolution, for radiation and $\beta = 0.04$. The velocities are binned in 10 bins with uniform width. The color code is analogous to that of Fig. 10: blue corresponds to strings that do not interact with other strings before the next time step (flow), green is for strings which merge with other segments and yellow is for strings that disappear before the next time step because the segment collapses. Note that in the last time step of segment velocities we have no information whether the segments will flow, merge or collapse, so we just choose to show them as flow segments. Note also that unlike in the segment case, where the velocity has been calculated *forward* and *backward*, for monopoles we have only calculated the velocity *forward*, and therefore we do not have information to compute the velocity at the last time step.

times and all lengths, specifically $v \sim 0.35$. The scatter is larger around short segments, mainly because there are many more short than long segments. This is rather clear in the *flow* case, because there are more segments in this case than in the *merge* case; though we could say that this is generic for all cases.

This information can be shown also, maybe in a more compact way, in the form of histograms; these are in some sense the marginalized distribution functions of Figs. 8 and 9. For example, in the top panel of Fig. 10 we show the distribution of the lengths divided by time of the number of segments during the evolution for $\beta = 0.04$ in radiation domination. The segments are binned in 10 bins with uniform width, and we show together three cases: in blue we depict the segments which are *flowing*, in green the ones that are *merging* and in yellow we show the segments that *collapse*. Remember that those behaviors are determined only for the reported time step, i.e., a segment that *flows* at a given time-step can be *merging* on the next one. Note that in the last time step we have no information whether the segments will flow, merge or collapse, so we just choose to show them as flow segments. In the bottom panel we show the same histogram, but instead of showing the number of segments in each bin, we also account for the length-per-time of the strings; we add up the total length-per-time of the segments in each bin, and show that length-per-time in each bin.

The histograms with information about velocities can be seen in Fig. 11. The top panel shows the distribution of number of segments per velocity (for $\beta = 0.04$ in radiation domination, as before). The bottom panel shows the corresponding distribution for monopoles. These velocities are computed instantaneously, i.e., they are the velocities obtained by measuring the distance traveled by monopoles between two adjacent time steps (not by averaging over the whole history of the monopole). Note that there is no

information for time $\tau = 256$ because we have no “next time step” to compute the velocity with. The color code is the same as in the previous figure: in blue we depict the strings (segments or monopoles, correspondingly) which are *flowing*, in green the ones that are *merging* and in yellow we show the strings that *collapse* before the next time step.

The histograms are consistent with a scaling behavior of the segment distributions. The length distributions show that, logically, only short segments collapse before the next time step. However, mergings happen for all string lengths. There is no preference on the velocity of the collapsing segments: there are collapsing segments with a wide variety of velocities. The velocities follow approximate Gaussian distributions, centered in the average velocity, corresponding to the velocities of Table III.

The monopole velocity histograms also show a (wider) Gaussian-like distribution, with instances in which monopoles approach $v = 1$ (though this may be an artefact of our algorithms, as explained in Sec. IV C). It is easy to convince oneself that the averages of these velocities agree with the ones in Table III. It is worth noticing that the velocities of monopoles that are about to annihilate are rather high: most of the times they are higher than the average velocities of the segments. Therefore, even though the segment velocities for collapsing segments show no evidence of a bias, the monopole velocities do. This could be understood by realizing that the monopole velocity only takes into account the string-ends, whereas in the segment velocity, the whole segment contributes, and apparently only the segment ends get high velocities in collapse.

Finally, it is interesting to investigate what the pattern of merging of different segments is. In order to do so, we chose one of the largest segments at the last time step for $\beta = 0.04$, and traced back its history to see what were its “constituents.” This information can be found in Fig. 12, in

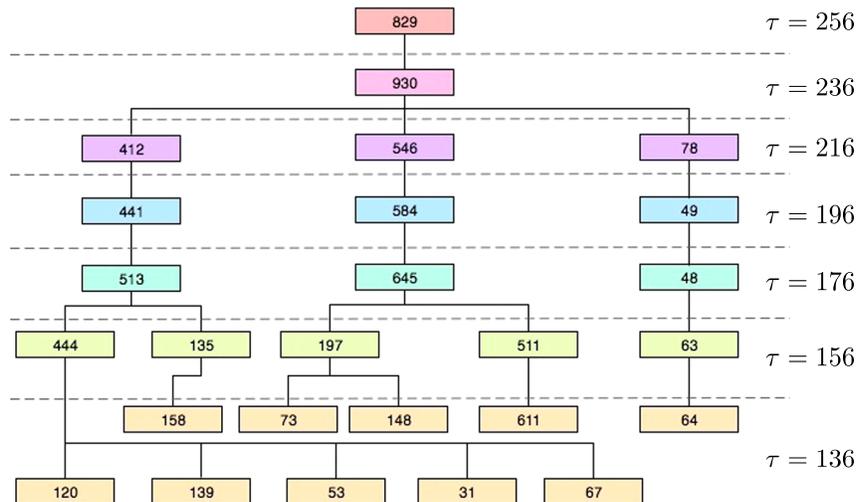


FIG. 12. The “family tree” of a segment for $\beta = 0.04$ in radiation. The number inside the box denotes the length of each segment, and time runs upwards.

the form of a “family tree.” This shows that at very early times, many mergings happen. This does not mean that many segments have joined at the same time; most likely the segments have been joining by pairs, but our choice of time steps to measure the network is too coarse to distinguish all these mergings. The sum of the length of the constituents does not match the final segment; clearly, this is because segments can grow or shrink in their evolution. Also, there are some segments that remain quite solitary for most of their life. This exercise of following the family history of a segment highlights once more the complicated dynamics of semilocal string networks.

VI. CONCLUSIONS AND DISCUSSION

In this work we have further investigated the evolution of networks of semilocal strings using field theoretical simulations. We have estimated the length and velocity of the semilocal strings (including the string ends, which can be understood as global monopoles) using different estimators, coming from both direct field theory based diagnostics and from identifications of the position of the strings. As well as characterizing the network of defects, these measures are indispensable to obtain a VOS-type effective model for semilocal strings. Before this work, the velocities of semilocal strings (both as a network and as individual segments, as well as the monopoles) were unknown.

The VOS-type length estimators for string segments and monopoles (14) can be found in Table II. These values had already been computed in a previous work [36] using a different technique. In [36] the semilocal string segments were defined as collections of adjacent points with a magnetic field higher than some threshold, and that volume of points was divided by an estimate of the cross section of the strings to get a length. In this work we repeated this analysis and compared it to the new length estimator, which is based on following the points with windings in the simulation. We found that these two length estimators do not match, and that there is roughly a factor 1.5 difference between the two length estimators. We identified the source of error as a numerical error on the estimation of the string cross-section, and therefore the values for the length estimators that we obtain in the present work can be considered as corrections to the ones in [36]. However, the differences obtained lie within the $1\text{-}\sigma$ uncertainties. It should also be noted that the length corresponding to the monopoles $\gamma_{\mathcal{N}}$ is roughly equivalent in both works, since the number of segments (monopoles) does not depend on the error corresponding to the cross-section of the strings.

One big advantage of being able to characterize the semilocal strings by the points (plaquettes) with winding is that we get a one-dimensional description for the position of the strings. This allows us to follow string segments and monopoles throughout the simulation, map their history, and get an estimate on the velocity. Table III shows the values of the network velocities for segments and

monopoles in the radiation and matter eras, for all values of β under study. The velocities for radiation domination are higher than the corresponding velocities in matter domination, for all β . It is worth pointing out that there is little dependence on β for these numbers, especially for the velocity of the segments.

The velocities can also be estimated using field-theoretical estimators of Eq. (11), and the result can be found in Table IV. The velocities obtained with both local-field estimators agree with each other, and as in the previous case, the velocities in radiation are higher than in matter. These values obtained from the field-estimators are somewhat different from the ones obtained from the positions of the defects. In fact, the field estimators do not distinguish between strings and monopoles, and they give a single number for the whole network. Unsurprisingly, in all cases the values lie in between the velocities of strings and monopoles.

Contrary to the velocities obtained from the positions, the values in Table IV show a trend with β . One cannot but speculate that this difference is the result of several factors in the simulation box: the ratios of densities of strings and monopoles depend on β (for lower β strings are longer), the tension of the strings (for lower β strings are lighter), the tendency of strings to collapse or merge... Moreover, it has been noted before that the estimation of defect velocities using directly the position of the defects underestimates the velocities [26,45,52], and the reason for this is not completely clear.

Even though there are some discrepancies in the values of the velocities, one huge advantage of being able to follow the positions of the strings is that we can obtain the history and the velocities for individual segments. Figures 8 and 9 show the scatter plot of segment length versus segment velocity, for cases when the segments do not interact with other segments in the following time step, and for cases where the segments merge with other segments, respectively.

It was somewhat surprising to see that the values of the velocities were scattered around a central value for all values of the length of the strings. In other words, we found no correlation between the length and the velocity of the segments. This is a hint that the VOS effective model may need to be revisited to take this into account, also in view on the recent revision of the VOS model for monopoles in [35]. Moreover, one could have wondered whether it is more preferable for short segments to collapse and for longer ones to merge. These plots show that mergings happen for all segment sizes, not only for longer ones.

The histograms where the length of the strings per bin are depicted (Fig. 10 bottom) show that actually most of the string length is not in the shortest segments; there is more length in the second and/or third bins. This histogram also shows more clearly that there are very long segments which do merge with other ones. This interesting behavior is also shown in the family tree-like figure Fig. 12. There we can see many mergings at the beginning of the simulation, then

some quiet period, before mergings happen also toward the end of the simulation. The lengths of the segments that take part in mergings are very varied, and there is no correlation between segment size and their future behavior.

Monopole velocities show, however, that segments that are going to collapse do it generally at somewhat higher than average velocities. The distribution is not so clearly Gaussian; there is more noise, and some monopoles are very fast. One has to be cautious here and remember all the caveats we have to overcome to obtain velocities of segments. The instances where monopole velocities were very close to unity (and sometimes seemingly faster than that) are due to the merging/collapsing of segments, and the shortcomings of our algorithm or interpretation of those events.

One of the most challenging aspects in this work has been, actually, the correct estimation of the velocity of segments and monopoles close to mergings. The naïve estimators of velocity gave instances of extremely high velocities. After we corrected for the obvious caveats (such as trying to get the velocity of a segment that had collapsed by trying to find the collapsed segment in the next time step and obviously failing) many of those very fast segments disappeared. But there were still some cases where the velocity was too high, higher than what one would naively expect from the causality of the field equations, and from what the field-estimators for velocities were showing.

The reason for this apparently superluminal behavior is that, close to segment mergings, if the field configuration is favourable, new string segments appear “out of the blue.” This is not a case of a monopole moving toward an antimonopole and creating string as it evolves; it is rather a new segment appearing, and giving the illusion of a very fast movement of the monopole/segment (in some sense, the far end of the string is a new monopole altogether, and the *old* monopole has annihilated with the newly formed antimonopole). This behavior is completely consistent with causal dynamics; it is the effect of the magnetic field accreting to form the new string segment. It has been studied in two-dimensional simulations in [55,57] and is also seen in the case of a global monopole whose cores are pinned down while letting the radial field gradients bunch into a string-like region that subsequently decays and disappears, taking the monopole with it [58–60].

Some of our conclusions are expected to be relevant for the VOS analytic model for semilocal strings. In particular, our results suggest that—unlike in the Abelian Higgs case, where smaller loops are faster—the velocity of semilocal segments is relatively insensitive to the size of the segments. Also, if we assume that network velocities are subluminal, the effective model has to account for the possibility of segments growing out of thin air, because otherwise the speed at which segments merge and collapse may be underestimated.

This work on the characterization of semilocal strings has highlighted even more how rich the dynamics of this model is, and how complicated the life of a semilocal network can be.

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