Solar Flare Heating with Turbulent Suppression of Thermal Conduction

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Abstract

During solar flares, plasma is typically heated to very high temperatures, and the resulting redistribution of energy via thermal conduction is a primary mechanism transporting energy throughout the flaring solar atmosphere. The thermal flux is usually modeled using Spitzer’s theory, which is based on local Coulomb collisions between the electrons carrying the thermal flux and those in the background. However, often during flares, temperature gradients become sufficiently steep that the collisional mean free path exceeds the temperature-gradient scale size, so that thermal conduction becomes inherently nonlocal. Further, turbulent angular scattering, which is detectable in nonthermal widths of atomic emission lines, can also act to increase the collision frequency and thus suppress the heat flux. Recent work by Emslie & Bian extended Spitzer’s theory of thermal conduction to account for both nonlocality and turbulent suppression. We have implemented their theoretical expression for the heat flux into the RADYN flare-modeling code and performed a parameter study to understand how the resulting changes in thermal conduction affect the flare dynamics and hence the radiation produced. We find that models with reduced heat fluxes predict slower bulk flows, less intense line emission, and longer cooling times. By comparing the features of atomic emission lines predicted by the models with Doppler velocities and nonthermal line widths deduced from a particular flare observation, we find that models with suppression factors between 0.3 and 0.5 relative to the Spitzer value best reproduce the observed Doppler velocities across emission lines forming over a wide range of temperatures. Interestingly, the model that best matches the observed nonthermal line widths has a kappa-type velocity distribution function.

Unified Astronomy Thesaurus concepts: Solar flares (1496); Solar chromosphere (1479); Solar corona (1483); Radiative transfer simulations (1967); Solar ultraviolet emission (1533); Hydrodynamical simulations (767)

1. Introduction

The flare component of a solar eruptive event rapidly heats the plasma in the Sun’s atmosphere to temperatures often greater than 20 MK (e.g., Fletcher et al. 2011; Caspi et al. 2014; McTiernan et al. 2019). The hot, flaring corona is thermally connected to the cooler chromosphere (≤20 kK) through a relatively narrow transition region with a steep temperature gradient, and the thermal conductive flux associated with such high temperature gradients is a primary mechanism for transporting energy throughout flaring loops (e.g., Aschwanden & Alexander 2001; Ryan et al. 2013; Warmuth & Mann 2016, 2020, and references therein).

When the electron-ion collisional mean free path $\lambda_{ei}$ is long compared to the temperature-gradient scale size $L$ (i.e., the Knudsen number $\lambda_{ei}/L \gg 1$), then thermal conduction becomes an inherently nonlocal process. In this regime, Campbell (1984) has demonstrated that thermal conduction is reduced relative to the Spitzer (1962) formalism. Additionally, Bian et al. (2016) have shown that the turbulent angular scattering of thermal electrons further reduces the thermal conductivity, and since thermal conduction is a main driver of flare dynamics, the overall evolution of temperatures, densities, and bulk velocities within the loop are all affected by the presence of turbulence.

The presence of turbulence is inferred from the broadening of atomic spectral line profiles and has been detected during flares in numerous previous studies (e.g., Landi et al. 2003; Del Zanna et al. 2006; Milligan 2011; Young et al. 2015; Jeffrey et al. 2016; Polito et al. 2018; Reeves et al. 2020, and references therein). Atomic emission line profiles are therefore sensitive to turbulence both directly (through nonthermal broadening) and indirectly, through the hydrodynamic evolution of flare loops driven in part by a turbulence-modified conduction term. This makes such line profiles ideal diagnostics for studying the role of turbulence in flares.

Emslie & Bian (2018, hereafter EB18) have developed a new formalism describing the suppression of thermal conduction that includes both nonlocal and turbulent effects in a single kernel that must be convolved with the local Spitzer heat flux to obtain the actual heat flux. In this work, we apply this approach to study the suppression of thermal conduction by both nonlocality and turbulence, by modeling the radiative hydrodynamic evolution of flare loops to several parameterizations of the turbulence spectrum. In Section 2, we present our methodology for performing this parameter study. These models can then be used to predict the emission profiles in many atomic lines, and in Section 3, we compare the predicted Doppler and nonthermal velocities with those observed during a particular flare. Finally in Section 4, we summarize these results and present our conclusions.

2. Methodology

We model the hydrodynamic response to the heating produced from an injection of nonthermal electrons into a
loop using the RADYN code (Carlsson & Stein 1992, 1995, 1997; Allred et al. 2005, 2015) combined with the FP particle transport code (Allred et al. 2020). RADYN has been used extensively in previous studies of flare heating (e.g., Abbett & Hawley 1999; Allred et al. 2005; Kennedy et al. 2015; Kowalski et al. 2015, 2017, 2022; Kowalski & Allred 2018; Kerr et al. 2016, 2019, 2020, 2021; Rubio da Costa et al. 2016; Simões et al. 2017; Brown et al. 2018; Graham et al. 2020), and its computational method is described in those works. Briefly, it solves the 1D equations of hydrodynamics coupled with nonequilibrium atomic level population equations for six levels of H, nine levels of He I and He II, and six levels of Ca II on an adaptive grid. The radiative transfer equation is solved using a 1.5D plane parallel approximation. A key feature of RADYN is its ability to model optically thick, non-LTE lines and continua for many transitions that are important for energy transport in the chromosphere, where much of the flare energy is deposited.

For this study, we have modified the expression that RADYN uses to compute the thermal conduction to incorporate the formalism for nonlocal and turbulent conduction presented in EB18. First the local Spitzer (1962) heat flux, $F_S$, is calculated as usual:

$$ F_S = -\kappa_e T^{5/2} \frac{dT}{dz}, $$

where $\kappa_e = 1.1 \times 10^{-6}$ erg K$^{-7/2}$ cm$^{-1}$ s$^{-1}$, $T$ is the plasma temperature (K), and $z$ is the distance (cm) along the loop, measured upward from the photosphere. $F_S$ is then convolved with a kernel, $Q$, that accounts for both nonlocality and turbulence, as derived in EB18. To model the effect of turbulence on thermal electron transport, EB18 assumed the velocity dependence of the turbulent mean free path, $\lambda_T(v)$, has the form

$$ \lambda_T(v) = \lambda_{T0} \left( \frac{v}{v_{Te}} \right)^\alpha = \frac{\lambda_{ei}}{R} \left( \frac{v}{v_{Te}} \right)^\alpha, $$

with $R = \lambda_{ei}/\lambda_{T0}$. Here

$$ v_{Te} = \sqrt{\frac{2kT}{m_e}} $$

is the electron thermal speed, and $\lambda_{ei}$ is the electron collisional mean free path at the thermal speed:

$$ \lambda_{ei} = \frac{m_e^2 v_{Te}^4}{4 \pi e^4 \ln \Lambda n}, $$

where $k$ (1.38 $\times$ $10^{-16}$ erg K$^{-1}$) is the Boltzmann constant, $e$ (4.8 $\times$ $10^{-10}$ esu) is the electronic charge, $n$ (cm$^{-3}$) is the background electron density, $m_e$ (9.1 $\times$ $10^{-28}$ g) is the electron mass, and in $\Lambda(\sim 20)$ is the Coulomb logarithm.

As shown by EB18, the form of the heat flux in the presence of nonlocal effects and/or turbulence is given by the convolution

$$ F_i(z) = \int Q(z - z'; R, \alpha) F_S(z') dz', $$

where the integral is done over the full-loop length, and the kernel function $Q$ is given by (EB18, their Equation (27))

$$ Q(z; R, \alpha) = \frac{\sqrt{45}}{60 \lambda_{ei}} \int_0^\infty x^5 \left( x^2 - \frac{5}{2} \right) \exp \left[ -\left( x^2 + \frac{45}{x^4 (1 + Rx^{4-\alpha})} \right) \frac{z}{\lambda_{ei}} \right] dx. $$

$Q$ is thus parameterized by $R$ and $\alpha$, and is plotted in Figure 1 for several values of $R$ and $\alpha$ that have been used in this study. We note that RADYN models half-loops, and by reflecting the RADYN half-loop about the apex, we make the assumption that the full-loop is symmetric.

The initial (preflare) state of the loop model has an apex temperature and electron density of 3.3 MK and 7.5 $\times$ 10$^9$ cm$^{-3}$, respectively. The temperature and density as a function of distance are plotted as $t = 0$ s (black line) in Figure 2. This state is obtained by evolving the radiative hydrodynamic equations with a coronal heating term as a source of energy at the top of the loop until a state of hydrostatic equilibrium is obtained, as described in Allred et al. (2015). In doing this relaxation to equilibrium, the unsuppressed Spitzer heat flux is used, so all models presented in Section 3 (listed in Table 1) have the same initial state.

We then inject a beam of nonthermal electrons at the apex of a model flux loop and track their transport until their eventual thermalization in the footpoints, typically in the chromosphere. For each case studied here, we inject the same nonthermal electron distribution, which is characterized by a power law with cutoff energy, $E_c = 12$ keV, spectral index, $\delta = 7.6$, and energy flux $F = 5 \times 10^{10}$ erg cm$^{-2}$ s$^{-1}$. This electron beam distribution was chosen to correspond to that which was inferred at the peak of a C-class flare studied by Milligan & Dennis (2009, hereafter M09) and Milligan (2011, hereafter M11) and will aid in comparing their observations with our model predictions presented in the subsequent section.

![Figure 1. The convolution kernel, $Q$ (Equation (6)), for several values of $R$ and $\alpha$.](image-url)
In each case, the beam is injected for $\Delta t = 20\text{ s}$ as estimated from observed impulsive rise times.

The transport of injected electrons is modeled using FP, which we have configured to run in a 1D mode, meaning all electrons are injected along the magnetic-field line axis and pitch-angle diffusion due to Coulomb collisions being neglected. As a test, we have run the cases in which the pitch-angle diffusion is included and found that it makes little difference to the beam heating rate, which is the important quantity for these simulations. We recognize that the diffusive scattering of the nonthermal electrons by turbulence could alter their energy deposition rate, especially if turbulent scattering is important (see Emslie et al. 2018). However, in order to highlight the effect of turbulence on the conductive term, here we do not consider its possible effect on the nonthermal electron transport, reserving such a study to a future work. FP numerically solves the Fokker–Planck equation including Coulomb collisions and ohmic losses associated with the return-current electric field, on an energy grid with 300 logarithmically spaced cells ranging from 0.1 keV to 10 MeV, using the method described in Allred et al. (2020). As configured, FP does not include the effects of runaway electrons, accelerated out of the ambient plasma by the return-current electric field, on the transport of beam electrons. These can be important when the electron beam flux is high.

![Figure 2](image_url)

**Figure 2.** The temporal evolution of the temperature (top row), electron density (middle row), and gas velocity (bottom row) for the Spitzer (left column), R1.0A0.9 (middle column), and R1.0A4.0 (right column) simulations. The line color represents the time from initial beam injection as indicated from the color bar.

| Label      | $R$  | $\alpha$ | $s$ | Period (s) | $\Delta t_T$ (s) | $\Delta t_n$ (s) |
|------------|------|----------|-----|------------|------------------|------------------|
| Spitzer    | N/A  | 1.0      | 1.00| 55.8       | 402              | 688              |
| R1.0A0.9   | 1.0  | 0.9      | 0.05| 34.2       | 1556             | 2117             |
| R1.0A1.7   | 1.0  | 1.7      | 0.10| 39.0       | 1143             | 1625             |
| R1.0A4.0   | 1.0  | 4.0      | 0.50| 49.3       | 552              | 884              |
| R2.0A4.0   | 2.0  | 4.0      | 0.33| 46.4       | 661              | 1027             |
| R10.0A9.0  | 10.0 | 9.0      | 0.90| 51.3       | 422              | 722              |
| R13.5A6.0  | 13.5 | 6.0      | 0.33| 46.3       | 662              | 1030             |

In each case, the beam is injected for $\Delta t = 20\text{ s}$ as estimated from observed impulsive rise times.

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but in this work, the flux is sufficiently low that runaways likely only comprise a small fraction of the return current (Alaoui et al. 2021). Turbulence, which causes suppressed thermal conduction, also produces increased electrical resistivity, causing larger return-current electric fields, but also a corresponding increase in the Dreicer field (Dreicer 1959). Thus, we expect that the fraction of runaways in the return current is not directly affected by the level of turbulence present.

3. Modeling of the Atmospheric Response

To study how the suppressed heat flux affects the evolution of flaring loops, we have chosen six sets of the parameters $R$ and $\alpha$, and performed flare simulations for each. These parameter sets are listed in Table 1, where the first column indicates the labels by which we refer to the models. For comparison, we also have performed a simulation of the unsuppressed case that implements thermal conduction using the standard Spitzer heat flux (Equation (1)), and it is labeled “Spitzer.” To best compare with the Spitzer theory, our “Spitzer” model simply implements the heat flux using Equation (1) and does not limit it by the electron free-streaming rate as was done in most of the previous RADYN simulations. The conductive flux suppression factor, $s$, is computed by integrating $Q\lambda_e$ over $z/\lambda_e$ (Figure 1), and is the ratio $F_c/F_S$ that would be obtained if $F_S$ does not appreciably vary over $z$, that is, when nonlocal effects are small. Of course, in these simulations, nonlocality can be important, and the actual conduction suppression can deviate from $s$.

The particular values of $R$ and $\alpha$ listed in Table 1 were chosen to span a wide range of thermal conduction suppression, with $s$ ranging from 0.05 to 1. The most suppressed models (R1.0A0.9 and R1.0A1.7) have narrow kernels (see Figure 1) and demonstrate cases where turbulence dominates over nonlocality. The opposite extreme, where nonlocality dominates over turbulence, is demonstrated with the model R10.0A9.0, which has a relatively broad kernel. We have chosen two parameter sets that have $\alpha = 4.0$ (R1.0A4.0 and R2.0A4.0), since these have the interesting property of producing kappa-type velocity distribution functions (see Equation (12) and the discussion in Section 3.2). These have moderate suppression, with $s$ equal to 0.5 and 0.33, respectively. Finally, we have chosen model R13.5A6.0, which has a similar suppression factor, $s = 0.33$, as the R2.0A4.0 model, but $\alpha = 6$ produces very broadened Maxwellian velocity distributions (again see Equation (12)).

3.1. Thermal Evolution

Figure 2 shows the evolution of the temperature, electron density, and gas velocity for the first 20 s of the R1.0A0.9, R1.0A4.0, and Spitzer simulations. We chose to show these three models because they represent the most, middle, and least suppressed cases. The R1.0A0.9 and R1.0A4.0 simulations reach hotter temperatures but have lower coronal densities than the Spitzer simulation. This is because the reduced heat flux limits how quickly heat can be moved from the corona into the chromosphere, causing it to bottle-up in the corona, and limiting the responding rate of chromospheric evaporation. The transition region forms at a height where the density is such that the radiative losses balance the heat flux divergence. A reduced heat flux, therefore, causes the transition region to form higher in the loop. The flaring transition region (at $t = 20$ s) is at 0.77 Mm, 0.79 Mm, and 0.87 Mm, for the Spitzer, R1.0A4.0, and R1.0A0.9 simulations, respectively.

In Figure 3, we plot the temporal evolution of the apex temperature and density for each of the flare simulations. Oscillations are apparent in each case, and their periods are written in the fifth column of Table 1. Models with more reduced heat flux have shorter oscillation periods. Since their coronal temperatures are hotter, they have faster sound speeds, so the pulses travel through the loop in shorter times.

We define the temperature ($\Delta T_T$) and density ($\Delta n$) relaxation times as the time intervals required for the model looptops to cool and drain to their preflare values. These are listed in the sixth and seventh columns of Table 1. More-suppressed models have longer relaxation times. As discussed in EB18, the turbulent suppression of conduction may aid in explaining the longer-than-predicted cooling times observed for numerous flares (Ryan et al. 2013). The temperature and density relaxation times as a function of $s$ are well fitted by power laws of the suppression factor $s$; $\Delta T_T \propto s^{-\beta_T}$, and $\Delta n \propto s^{-\beta_n}$, with indices $\beta_T = 0.46$ and $\beta_n = 0.38$, respectively, as shown in Figure 4.
3.2. Spectral Line Profiles

Using RAYDIN’s predicted temporal evolution of the temperatures and electron and hydrogen densities combined with the atomic line contribution functions obtained from the CHIANTI package (v.10.0.1; Dere et al. 1997; Del Zanna et al. 2021), we have constructed the line intensities along our model loop for 14 optically thin transition region and coronal atomic lines presented in MD09 (listed in their Table 1), with the exception of the He II 256 Å line, which is likely optically thick and not well suited to CHIANTI’s coronal approximation. We use CHIANTI’s ionization equilibrium table and the abundance table from Schmelz et al. (2012), but we note that, since our primary purpose in this section is to compare the profile shapes and Doppler velocities, the differences in abundance tables are not critical to our study.

In a thermal nonturbulent plasma, the velocity distribution function, \( f(v) \), is a Maxwellian with width given by the thermal velocity. With this assumption, the line profiles will be Gaussians. However, the presence of turbulence causes the velocity distribution function to deviate from a Maxwellian resulting in non-Gaussian line profiles, with widths that are typically wider than than the thermal width. Bian et al. (2014) solved for the velocity distribution function in a turbulent environment using a steady-state approximation to the Fokker–Planck equation. They found a distribution function of the form (their Equations (7)–(10)),

\[
 f(v) = A e^{-U(v)},
\]

with \( U(v) \) the solution of

\[
 \frac{dU}{dv} = \left( \frac{v_c^2}{2v} + \frac{v^2D_{\text{turb}}}{\Gamma} \right)^{-1},
\]

where \( \Gamma = \frac{v^3}{\lambda_e} \). The turbulent diffusion coefficient, \( D_{\text{turb}} \), is given by

\[
 D_{\text{turb}} = \frac{v^3}{\lambda_e^2} = \frac{v^{3-\alpha}v_e^\alpha}{\lambda_e^2},
\]

where we have used the expression for the turbulent mean free path in Equation (2). Substituting this into Equation (8) gives

\[
 \frac{dU}{dv} = \left( \frac{v_c^2}{2v} + R \frac{v^{5-\alpha}}{v_e^{4-\alpha}} \right)^{-1} = \frac{2v v_e^{4-\alpha}}{v_e^{6-\alpha} + 2R v^{6-\alpha}}.
\]

In terms of the dimensionless speed \( w = v/v_e \),

\[
 \frac{dU}{dw} = \frac{2w}{1 + 2Rw^{6-\alpha}}.
\]

Finally, we obtain the distribution function by integrating this and substituting into Equation (7)

\[
 f(w) = A_0 \exp \left[ -\int_0^w \frac{2dx}{1 + 2Rx^{6-\alpha}} \right],
\]

where the coefficient \( A_0 \) is chosen to normalize \( f(w) \) to the local density. From this form of \( f(w) \), we note the following:

1. In the no-turbulence limit, i.e., when \( R = 0 \), \( f(w) \) regains its thermal Maxwellian form, as expected;
2. For the case with \( \alpha = 6 \), \( f(w) \) is also a Maxwellian, but is now broadened by a factor of \( \sqrt{1 + 2R} \). This result follows from the fact that the collisional and turbulent terms appearing in the Fokker–Planck equation have the same velocity dependence (i.e., \( w^{-3} \)). Thus, as previously noted by Bian et al. (2014), the presence of a Maxwellian distribution function is not exclusive to the collisionally-only situation. Caution must therefore be observed in inferring the local temperature from the width of an observed Gaussian line profile;
3. For the case with \( \alpha = 4 \), \( D_{\text{turb}} \) has a velocity dependence of \( w^{-1} \). In this case, \( f(w) \) becomes a kappa distribution of the first kind, as discussed in Section 3 of Bian et al. (2014), with \( \kappa = 1/2R \);
4. For \( \alpha < 4 \), the dependence of \( D_{\text{turb}} \) on \( w \) is such that, as \( w \to \infty \), \( f(w) \) behaves as an exponential of a negative power of \( w \). For example, with \( \alpha = 2 \), \( f(w) \to A e^{l/(2Rw^2)} \), while for \( \alpha = 0 \), \( f(w) \to A e^{1/(4Rw^4)} \). For large \( w \), these approach an asymptotic value, \( A \), but this steady state is only reached as \( t \to \infty \) (Chavani & Lemou 2005). So, to normalize \( f(w) \) for \( \alpha < 4 \), we truncate \( f(w) \) at 10 Doppler widths and then subtract off the asymptotic value to ensure that the line profiles go to zero for large \( w \).

For a given set of \( R \) and \( \alpha \), the line profiles are obtained by numerically calculating the integral in Equation (12) and are normalized so that the integral over wavelength is the total line intensity. The line profiles are then Doppler shifted by the bulk velocity along the loop axis projected along an observer’s line of sight. MD09 measured the line-of-sight Doppler velocities for several lines in Hinode/EUV Imaging Spectrometer (EIS) pixels that corresponded to the location of the hard X-ray footpoint as observed by RHESSI (see their Figures 3–5). To better enable comparisons with this flare observation, we orient our model loop footpoint at the same location on the solar disk, which is \((X, Y) = (550'' , -94'')\). At the footpoint, we assume that the loop is oriented along the vertical axis. Then using the WCS conversion routine, WCS_CONV_HPC_HCC, the projection between an observer at Earth and the vertical at the footpoint is readily determined to be \( \cos(35.76) = 0.81 \). We use this factor as the line-of-sight projection of bulk flows along our model loop. The EIS observation of this flare used a slit with angular width of \( 2'' \), spectral instrumental full width at half maximum (\( W_{\text{inst}} \)) of approximately 0.056 Å (Harra et al. 2009), and a temporal integration time of 10 s, so to enable a comparison to the EIS observation, we bin the predicted emissions from our model loops into \( 2'' \) bins and integrate them over 10 s intervals. The modeled line profiles are convolved with a Gaussian with the width given by \( W_{\text{inst}} \) to simulate instrumental broadening.

We have produced the line profiles for the 14 lines listed in Table 2. As examples, we show the line profiles for the first 10 s interval of each of our simulations for the \( 0 \) \( 184 \) Å, \( Fe \) \( 195 \) Å, \( Fe \) \( XVI \) \( 263 \) Å, and \( Fe \) \( XXIV \) \( 192 \) Å lines in Figure 5. Since a reduced heat flux drives slower chromospheric evaporation with a less dense corona, for each of these lines, the models with more-suppressed heat flux are less intense and exhibit weaker Doppler shifts. For example, comparing our most suppressed model (R1.0A0.9) with the Spitzer model, the \( Fe \) \( XXIV \) \( 192 \) Å line has integrated intensities of \( 2.2 \times 10^3 \) and \( 5.0 \times 10^3 \) erg \( cm^{-2} \) s\(^{-1} \) sr\(^{-1} \) and Doppler shifts of \(-144\) and \(-328 \) km s\(^{-1} \), respectively.

Characteristics of the line profiles are summarized in Table 2. The third column is the ion formation temperature from CHIANTI v.10.0.1. These differ slightly for a few lines compared to M11, which used CHIANTI v.6.0.1. The fourth
Table 2

Doppler ($v$) and Nonthermal ($v_{NT}$) Velocities Measured in Several Lines from MD09/M11 and from Our Models

| Ion  | $\lambda_0$ (Å) | $T_{max}$ (MK) | MD09/M11 | Spitzer R1.0A0.9 | R1.0A1.7 | R1.0A4.0 | R2.0A4.0 | R10.0A9.0 | R13.5A6.0 |
|------|-----------------|---------------|----------|------------------|----------|----------|----------|----------|----------|
| O VI | 184.12          | 0.3           |          | $v$ | $v_{NT}$ | $v$ | $v_{NT}$ | $v$ | $v_{NT}$ | $v$ | $v_{NT}$ | $v$ | $v_{NT}$ |
| Mg VI| 268.99          | 0.4           |          | 15 | 15 | 2 | 25 | 4 | 30 | 12 | 46 | 10 | 59 |
| Mg VII| 280.74         | 0.6           |          | 11 | 16 | -1 | 28 | 0 | 32 | 7 | 49 | 6 | 64 |
| Fe VIII | 185.21       | 0.6           |          | 16 | 28 | -1 | 33 | 1 | 37 | 11 | 56 | 9 | 68 |
| Fe X | 184.54         | 1.1           |          | -2 | 22 | -9 | 33 | -9 | 35 | -2 | 56 | -3 | 72 |
| Fe XI | 188.22         | 1.3           |          | -5 | 33 | -6 | 49 | -5 | 48 | 4 | 71 | 10 | 96 |
| Fe XII | 195.12        | 1.6           |          | -1 | 43 | -5 | 52 | -3 | 51 | 5 | 83 | 7 | 101 |
| Fe XIII | 202.04       | 1.8           |          | -15 | 24 | -16 | 38 | -15 | 36 | -15 | 61 | -14 | 79 |
| Fe XIV | 274.20        | 2.0           |          | -22 | 24 | -17 | 35 | -16 | 33 | -20 | 57 | -19 | 74 |
| Fe XV | 284.16         | 2.2           |          | -31 | 28 | -21 | 35 | -21 | 32 | -28 | 61 | -27 | 76 |
| Fe XVI | 262.98        | 2.8           |          | -39 | 20 | -29 | 39 | -31 | 37 | -46 | 77 | -43 | 92 |
| Fe XVII | 269.42       | 5.6           |          | -69 | 18 | -50 | 44 | -59 | 49 | -90 | 102 | -84 | 121 |
| Fe XXIII | 263.77       | 14.1          |          | -230 | 32 | 122 | -292 | 55 | -132 | 31 | -158 | 28 | -242 | 108 |
| Fe XXIV | 192.03        | 17.8          |          | -257 | 28 | 105 | -328 | 31 | -144 | 10 | -172 | 20 | -273 | 112 |

Note.

Formation temperatures are from CHIANTI v.10.0.1 (Del Zanna et al. 2021) and differ slightly from those listed in M11, which used CHIANTI v.6.0.1.
and fifth columns are the Doppler speed (with uncertainties) and nonthermal velocities determined in MD09/M11 (see Table I of M11). The left and right columns under each model heading are the Doppler and nonthermal velocities, respectively. The Doppler shifts are calculated from the line profile centroid. The nonthermal line velocities are found by measuring the full width at half maximum, \(W\), and subtracting the contributions from thermal and instrumental broadening. Specifically, we measure the nonthermal velocity, \(v_{NT}\), using the expression

\[
v_{NT} = \sqrt{\frac{c^2}{4 \lambda_c^2 \ln 2} (W^2 - W_{ma}^2) - \frac{2kT_{max}}{m_i}},
\]

where \(c\) is the speed of light, \(\lambda_c\) is the line center wavelength, \(T_{max}\) is the temperature at the peak of the contribution function (from CHIANTI), and \(m_i\) is the ion mass.

### 3.3. Doppler Velocities

To better visualize how the Doppler velocities compare to the MD09/M11 observations, in the left panel of Figure 6, we show the Doppler velocities predicted from each of our simulations as a function of line formation temperature (see also Figure 5 of MD09). All of our simulations produce downflows in cooler lines (chromospheric condensations) and upflows in hotter lines (chromospheric evaporations). Similar to the MD09/M11 observations, the R1.0A4.0, R2.0A4.0, and R13.5A6 simulations predict the transition between upflows and downflows occurring between the Fe XII \(T_{max} = 1.6\) MK and Fe XIII \(T_{max} = 1.8\) MK lines. However, the simulations predict downflows that are significantly slower than the observations. For example, we find flows less than 16 km \(s^{-1}\) in the O VI line compared to the observed \((60 \pm 14)\) km \(s^{-1}\). The only exception is the Fe XI 188 Å line in the R13.5A6.0 model, which appears to have a faster downflow \((29 \text{ km s}^{-1})\) compared to the other models. However, that is a result of its very broad line profile merging with a neighboring (redward) Fe XI line and not a real redshift. The too slow
downflows could be an indication that the preflare model chromosphere is relatively overdense compared to the real Sun, or perhaps the energy flux density in the electron beam was underestimated. This is possible if the footpoint area is not fully resolved.

The speed of the chromospheric evaporation is more directly related to the heat flux. The models with low suppression, Spitzer and R10.0A9.0, predict upflows that are faster than observed in the Fe XVI and hotter lines. For the strongly suppressed models, the opposite is true. The R1.0A0.9 and R1.0A1.7 models predict −144 and −172 compared to the observed <−257 km s$^{-1}$ for the FeXXIV 192 Å. Interestingly, the moderately suppressed models, R1.0A4.0, R2.0A4.0, and R13.5R6.0, predict upflow velocities that are within the MD09/M11 uncertainties. Therefore, these have plausible levels of thermal conduction suppression.

### 3.4. Nonthermal Velocities

The nonthermal velocities listed in Table 2 are plotted in the right panel of Figure 6. Even though the Spitzer simulation has no turbulence, it shows a small level of excess broadening (seventh column of Table 2), resulting from the superposition of a distribution of Doppler shifts occurring within the temporal and spatial bins, as speculated by M11 and discussed in Polito et al. (2019) and Kerr et al. (2020) in the context of Fe XXI. But this small amount of broadening is only comparable to the M11 observations for the Fe XVI 263 Å and Fe XVII 269 Å lines. For other lines, especially the very hot Fe XXIII and Fe XXIV lines, the Spitzer simulation is unable to produce sufficiently broad line profiles. For example, M11 observed $v_{NT} = 122$ km s$^{-1}$ in Fe XXIII 264 Å line compared to 55 km s$^{-1}$ predicted by the Spitzer simulation. The highly suppressed models, R1.0A0.9 and R1.0A1.7, also predict line profiles that are much narrower than observed. The R13.5A6 model has a broadening factor of $\sqrt{1 + 2(13.5)} = 5.3$ and produces line profiles that are much broader than in M11, e.g., 313 km s$^{-1}$ in the Fe XXIII line. M11 did not report the uncertainties in the nonthermal line widths, but assuming that they are comparable to the reported Doppler velocity uncertainties, we find that R1.0A4.0, R2.0A4.0, and R10.0A9.0 produce nonthermal broadenings that are similar to the observations. Of these, R1.0A4.0 has the smallest squared residuals. As described in the previous section, this model also produces Doppler velocities that match well with the observations. It is, therefore, plausible that the nonthermal line widths and Doppler velocities observed by MD09/M11 indicate a moderate level of turbulence produced during that flare. Interestingly, in our best-fit model, since $\alpha = 4$, the velocity distribution function is characterized by a kappa distribution, with a rather low value of $\kappa = 1/2R = 1/2$. In the analysis of other flares, Polito et al. (2018) and Jeffery et al. (2016) also found kappa-distribution (with significantly higher values of $\kappa \approx (2 − 3)$) best-fit line profiles, and Polito et al. (2018) concluded that turbulence is a likely cause.

### 4. Summary and Conclusions

We have used the RADYN flare-modeling code to study how the effects of turbulence and nonlocality on conductive heat transport affect flare radiation hydrodynamics. Specifically, we have implemented the EB18 thermal conduction model into RADYN and performed a parameter study, varying $R$ and $\alpha$, to understand how the suppressed heat flux affects the speed of chromospheric evaporation and the timescales of flare cooling and draining. We have studied seven sets of parameters with suppression factor ranging from 0.05 to 1.0 (i.e., unsuppressed). We used RADYN to simulate an injection of a beam of nonthermal electrons and the subsequent heating. The same beam distribution was used for each simulation, and was chosen to match the power law inferred from a RHESSI observation of a C-class flare by MD09. Our key findings are summarized below:

1. The transition region forms at a lower density in more highly suppressed models, resulting in lower intensities in transition region and coronal spectral lines;
2. Models with more highly suppressed thermal conduction result in longer cooling ($\Delta T_\text{f}$) and draining ($\Delta T_\text{d}$) times. These are well fitted by power laws of the suppression factor $s$, $\Delta T_{f,d} \propto s^{-\beta_{f,d}}$ with indices $\beta_{f,d}$ equal to 0.46 and 0.38, respectively;
3. The flaring chromospheric evaporation front is slower for models with more highly suppressed thermal conduction.
The models R1.0A4.0, R2.0A4.0, and R13.5R6.0, which have $s = 0.5$, 0.33, and 0.33, respectively, produce upflow Doppler shifts in several lines that are comparable to observations of a C-class flare presented in MD09/M11.

4. Predicted nonthermal line widths, primarily due to turbulent broadening, from the R1.0A4.0, R2.0A4.0, and R10.0A9.0 models are comparable to the M11 observations;

5. Overall, the R1.0A4.0 model produces Doppler velocities and nonthermal line widths that best match the observations; this model has a moderate level of turbulent broadening, from the R1.0A4.0, R2.0A4.0, R13.5R6.0, and R10.0A9.0 models are comparable to the M11 observations.

This work represents a first step in understanding how turbulence affects the evolution of flare loops and the resulting emission. In future work, we will further improve upon our models by including the effects of turbulent diffusion on the transport of the nonthermal electrons and the effects of runaway electrons accelerated out of the ambient thermal distribution.

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