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2 *Geophysical Research Letters*

3 Supporting Information for

4 **Wildfire impact on environmental thermodynamics and severe convective storms**

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16 **Contents of this file**

17 Supplementary text

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19 **Introduction**

20 The supplementary text includes some detail description about model development,
21 WRF-Chem model configurations and simulations, the observational datasets used
22 for model evaluation, and the references cited in this document. Description of
23 Figures S1-S8 are shown in the corresponding figure caption.

24 **Supplementary text**

25 **Description of model development**

26 For the model development, we provide more details about biomass emissions, fire
27 location, timing, plume height, etc. The hourly biomass emission, active fire location, and
28 burning area were obtained from the Fire INventory from NCAR (FINN) model
29 (Wiedinmyer et al., 2011). The location and timing for the fires are identified globally
30 using the MODIS Thermal Anomalies Product. The Global Land Cover Characteristics
31 2000 dataset and the MODIS Vegetation Continuous Fields Product were used to
32 determine the burned area for each vegetation type based on percentage vegetative cover
33 in each 1 km² fire pixel. The FINN data is then interpolated to the host model grid. For grid
34 points with fire according to FINN, the sub-grid plume rise model is driven by the
35 environmental dynamics from the atmosphere model in WRF-Chem and the plume
36 dynamics are estimated based on fire information from FINN (Freitas et al., 2007). The
37 final height of the plume is then used in the source emission field of the host model to
38 determine the effective injection height where heat and aerosols emitted during the flaming
39 phase would be released and interact with the atmosphere circulation and transport. The
40 aerosol emission from the fire is added as described in Grell et al. (2011).

41 The sensible heat flux from the subgrid plume model is input to the atmosphere as
42 an additional source term in the equation for potential temperature θ , equal to the vertical
43 divergence of the heat flux,

$$44 \frac{d(\mu\theta)}{dt}(x, y, z) = R_\theta(\Phi) + \frac{\mu(x,y)\phi_h(x,y)}{\sigma\rho(x,y,z)} \frac{\partial}{\partial z} \exp\left(-\frac{z}{z_{ext}}\right) \quad (1)$$

45 where $\mu(x,y)$ is the hydrostatic component of the pressure differential of dry air between
46 the surface and the top of the domain, $R_\theta(\Phi)$ is the component of the source term
47 (commonly called “tendency” in WRF) in the atmospheric model thermodynamic
48 equation, σ is the specific heat of the air, $\rho(x,y,z)$ is the density, and z_{ext} is the heat
49 extinction depth (Mandel et al. 2011).

50 **Model configurations and simulations**

51 For the evaluation of the improved WRF-Chem model at 3-km resolution for
52 wildfires without pyroCb, we run the WRF-Chem simulations of wildfires over the central
53 United States from 1200 UTC 15 July to 1200 UTC 19 July 2016. The simulated domain
54 is shown in Fig S1a with 65 vertical levels. We used the Model for Simulating Aerosol
55 Interactions and Chemistry (MOSAIC) aerosol model with four bins (Zaveri et al., 2008).
56 The physics schemes applied in the simulation are the Unified Noah land surface scheme
57 (Chen and Dudhia, 2001), Yonsei University planetary boundary layer scheme (Hong et
58 al., 2006), the rapid radiative transfer model for general circulation model (RRTMG)
59 longwave and shortwave radiation schemes (Iacono et al., 2008), and Morrison two-
60 moment microphysics scheme (Morrison et al., 2005) with the hail option. The initial and
61 lateral boundary conditions for the meteorological fields were produced from the Rapid

62 Refresh (RAP) model that is comprised primarily of a numerical forecast model and an
63 analysis/assimilation system at 13-km resolution (Benjamin et al. 2016). The chemical
64 lateral boundary and initial conditions were created from the Modern-Era Retrospective
65 analysis for Research and Applications, Version 2 (MERRA-2, Gelaro et al., 2017). The
66 meteorological field was reinitialized every 30 hours with the RAP data. The anthropogenic
67 emission was from NEI-2011 emissions. The biogenic emission came from the Model of
68 Emissions of Gases and Aerosols from Nature (MEGAN) product (Guenther et al., 2006).
69 To evaluate the temperature profiles on 18 July 2016, we use the sounding data from the
70 National Oceanic and Atmospheric Administration / Earth System Research Laboratory
71 ((NOAA/ESRL) radiosonde database.

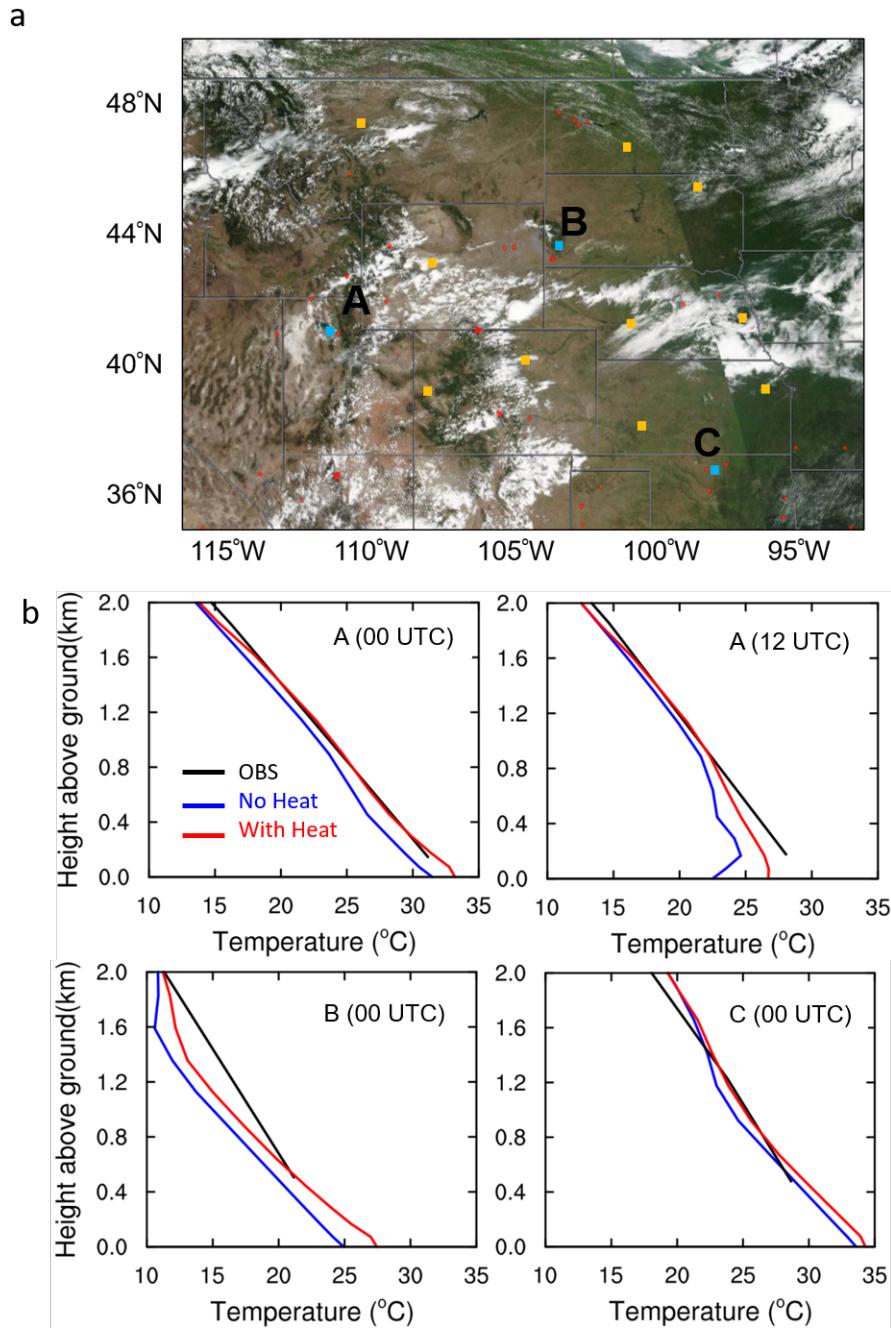
72 To evaluate Wildfire with the new plume model development, WRF-Chem-SFIRE
73 is run, which uses the similar model configuration as Wildfire. The inner fuel model for
74 fire in WRF-Chem-SFIRE uses a resolution of 50 meters that is 20 times finer than the
75 atmospheric model grid. The Anderson 13 fuel category data and high-resolution
76 topography data available at <http://www.landfire.gov> are used in SFIRE for estimating fire
77 behavior and spread.

78 Description of datasets used for model evaluation

79 The maximum hail sizes from the simulations are estimated using a physically-
80 based hail forecasting model (HAILCAST), which is online coupled with WRF-Chem
81 simulations (Adams-Selin and Ziegler, 2016). HAILCAST forecasts the maximum
82 expected hail diameter at the surface using updraft and microphysical information
83 produced by WRF-Chem. We incorporate the most updated HAILCAST version from
84 WRF v4.0 (Adams-Selin et al., 2018) into the WRF-Chem V3.9.1 for this study. The
85 prediction of lightning activity from model simulations is estimated with the utilization of
86 the lightning potential index (LPI) described in Yair et al. (2010). The smoke plume height
87 data digitized from the Multi-angle Imaging Spectro Radiometer (MISR) based on the
88 MISR INteractive eXplorer (MINX) software are utilized to evaluate the predicted plume
89 height (Nelson et al., 2014). To analyze and evaluate the thermodynamics before the
90 convection, sounding data from National Weather Service forecast office at Amarillo, TX
91 (KAMA) is used. The observed radar reflectivity and the radar-retrieved maximum
92 expected size of hail (MESH) data are from Gridded NEXRAD WSR-88D Radar data
93 (GridRad; Homeyer and Bowman, 2017) created at 5-min temporal intervals for this study.
94 The MESH data used in this study are developed from a newly-improved algorithm
95 (Murillo and Homeyer, 2019). The National Centers for Environmental
96 Prediction /Environmental Modeling Center ((NCEP/EMC) Stage IV Data is used as the
97 observation of precipitation with hourly output at 4-km resolution (Lin and Mitchell, 2005).
98 The lightning observation data are from National Lightning Detection Network (NLDN;
99 Cummins and Murphy, 2009).

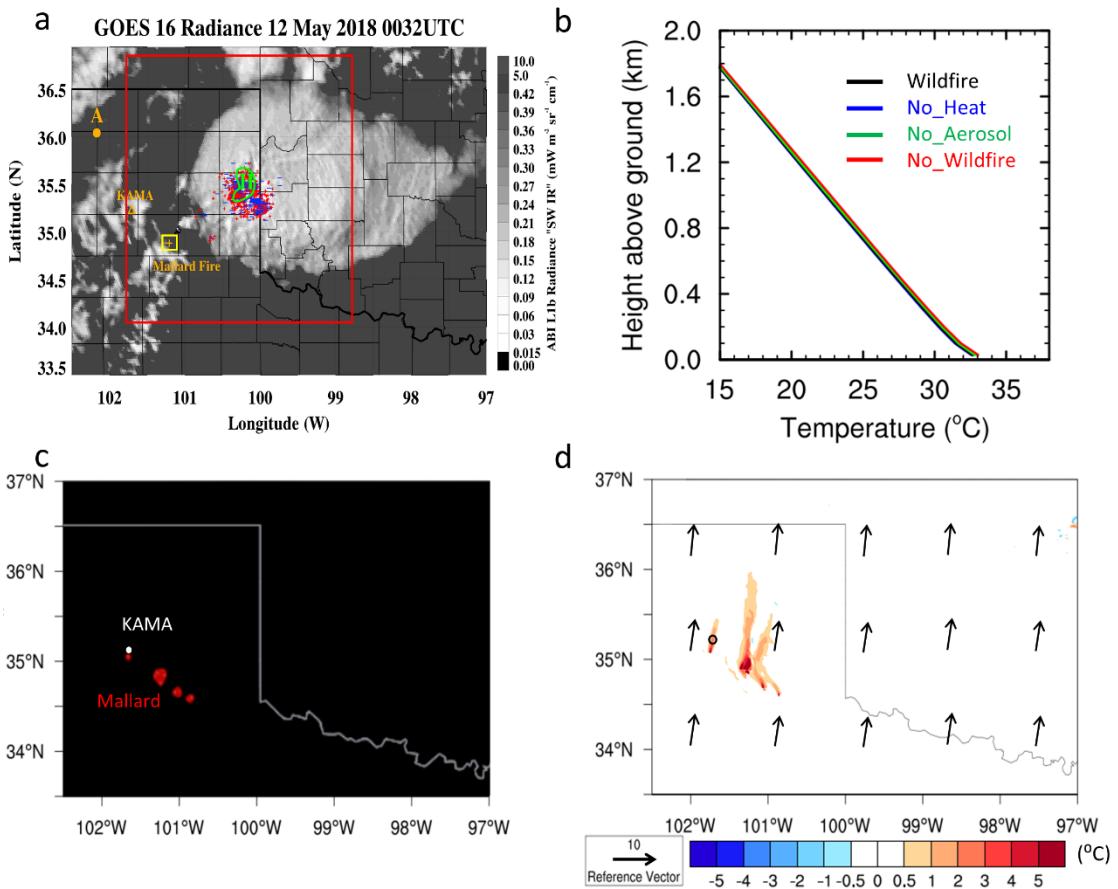
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- 133 **Figures S1 to S8**
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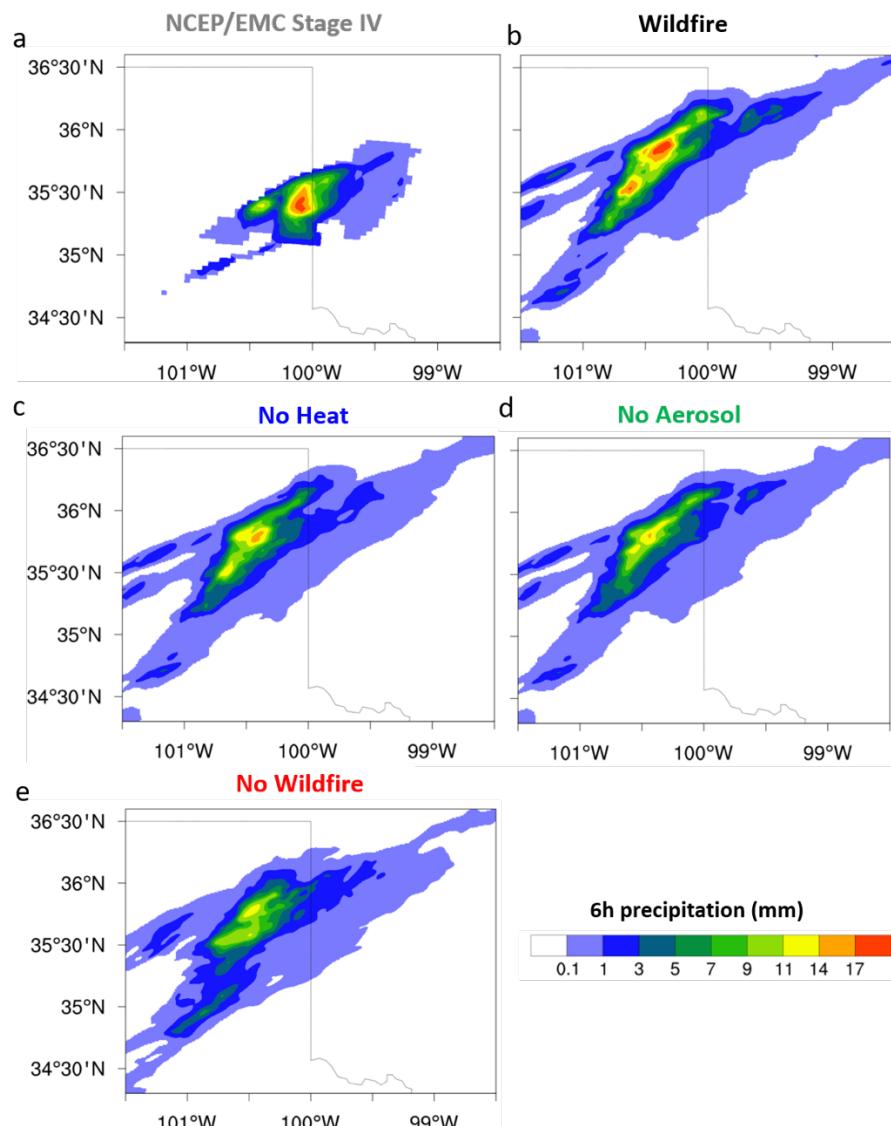
136 **Figure S1** (a) True Color image and Fires/Thermal Anomalies (red dots) from Suomi NPP/
 137 VIIRS and NOAA/ESRL Radiosonde stations (blue or yellow squares) in the simulated
 138 domain on 17 July 2016, (b) Temperature profiles from the observation (black), the
 139 simulation with the original WRF-Chem (blue) and the simulation from the improved
 140 WRF-Chem with heat flux accounted (red) for three sounding stations close to fires (blue
 141 squares in a) on 18 July 2016. Other stations did not have measurements obviously
 142 influenced by wildfires.



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144 **Figure S2** (a) The pyroCb observed from GOES-16 Band 7 (“shortwave window” Infrared)
 145 and the lightning flashes (marked as “+”) from the National Lightning Detection
 146 Network (NLDN) at 0032 UTC 12 May 2018, and the area of observed hail (green contour
 147 line) from the MESH data at 0030 UTC. The location of A is a site that was not influenced
 148 by wildfire and KAMA is a sounding site. The latitude and longitude ranges of (a) show
 149 the model simulation domain. The red box is the study domain for analysis of convection.
 150 The yellow box is for the analysis of temperature and moisture profiles over the Mallard
 151 fire area shown in Figure 1g, h. (b) Profile of temperature at 0000 UTC 11 May at the site
 152 A from the simulations. (c) Fire location detected by MODIS Thermal Anomalies Product
 153 on 10 May. (d) 2-m temperature anomaly (shaded) from No_Wildfire to Wildfire and 10-
 154 m wind (arrows) in Wildfire at 0000 UTC 11 May. KAMA sounding site is marked as
 155 circle.
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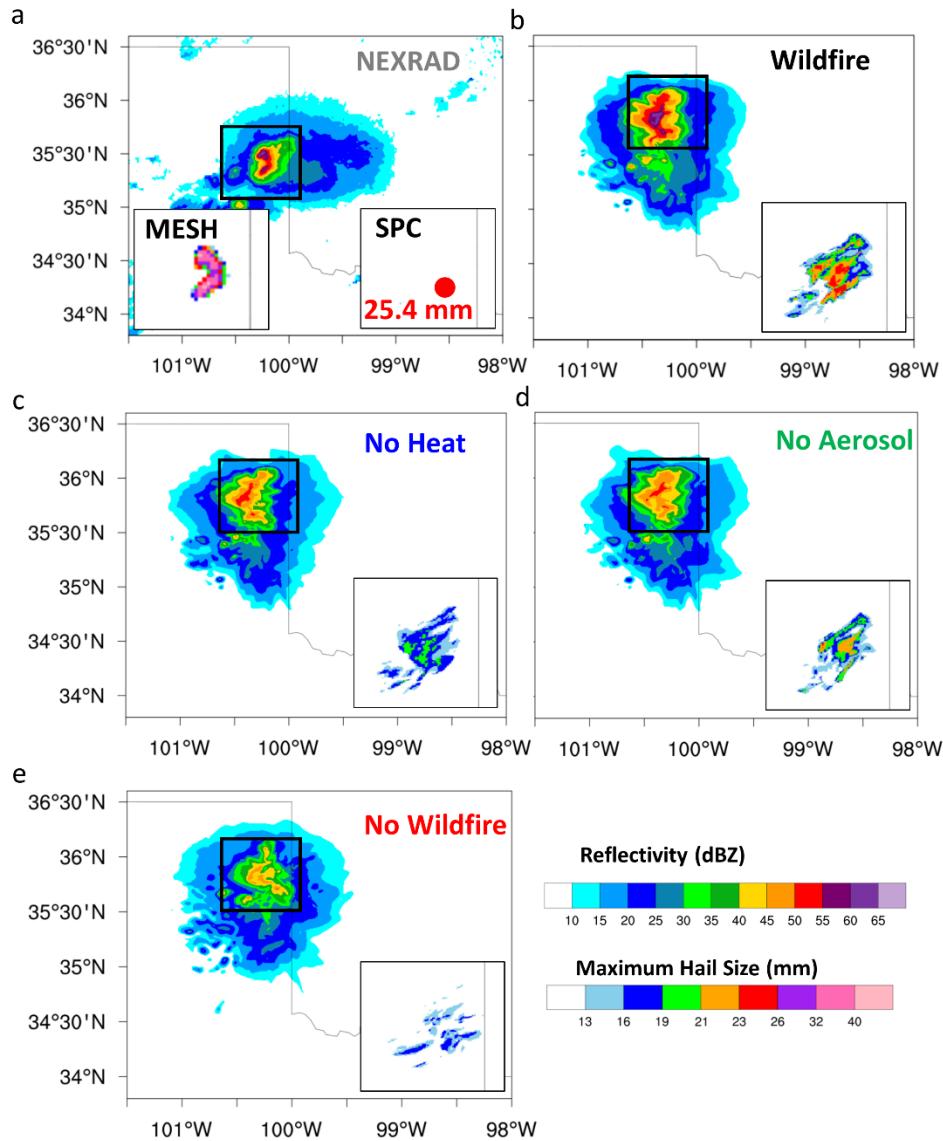
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159 **Figure S3** Accumulated precipitation from (a) NCEP/EMC Stage IV data, (b) Wildfire, (c)
160 No_Heat, (d) No_Aerosol, and (e) No_Wildfire over a 6-h time period from 2000 UTC 11
161 May to 0200 UTC 12 May.

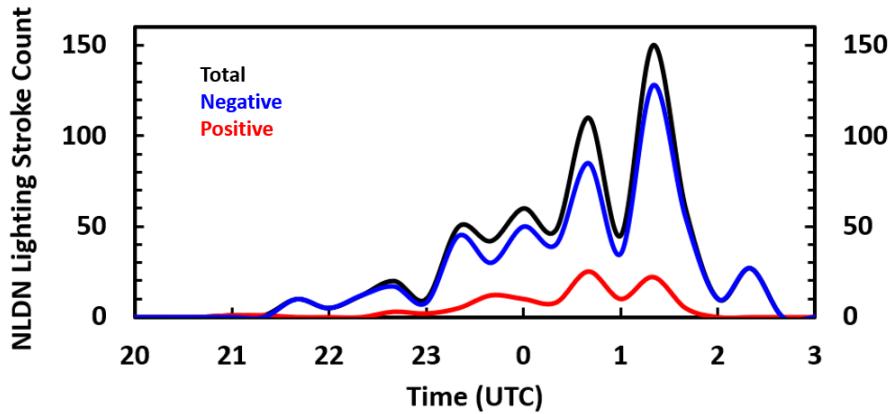
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164 **Figure S4** Composite reflectivity at the time when the maximum reflectivity is reached in
 165 temporal evolution from (a) NEXRAD at 0015 UTC 12 May, (b) Wildfire, (c) No_Heat,
 166 (d) No_Aerosol, (e) No_Wildfire at 2330 UTC 11 May. The corresponding maximum hail
 167 size is shown in the bottom-embedded small boxes for the black box region marked on the
 168 reflectivity plot. Both the SPC report and MESH data are shown on (a). The modeled
 169 results are from the HAILCAST estimation.

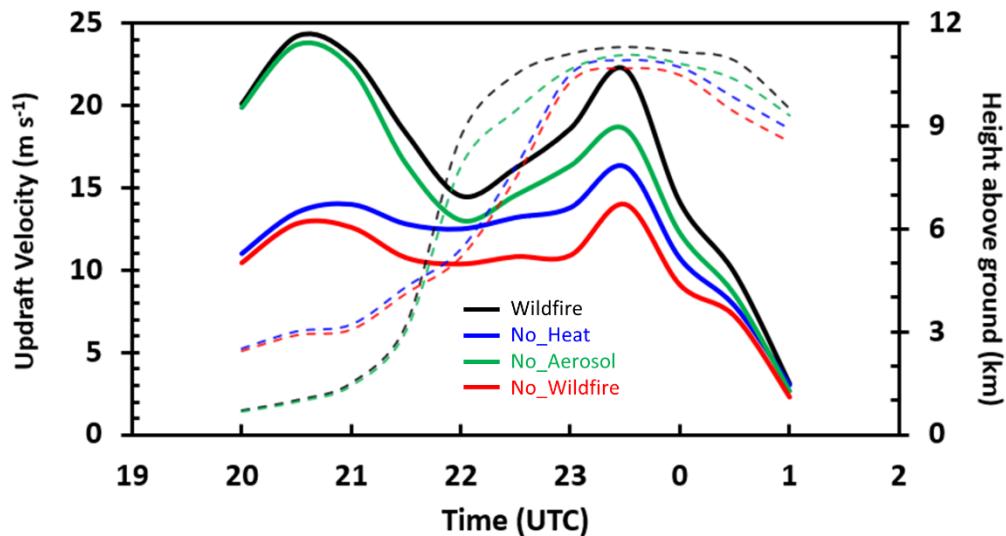
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172 **Figure S5** Time series of CG lightning stroke (flashes with all positive CG lightning
173 greater than or equal to 15 kA) from NLDN for the Mallard pyroCb (red for the positive
174 lightning; blue for the negative lightning, and black for total lightning).

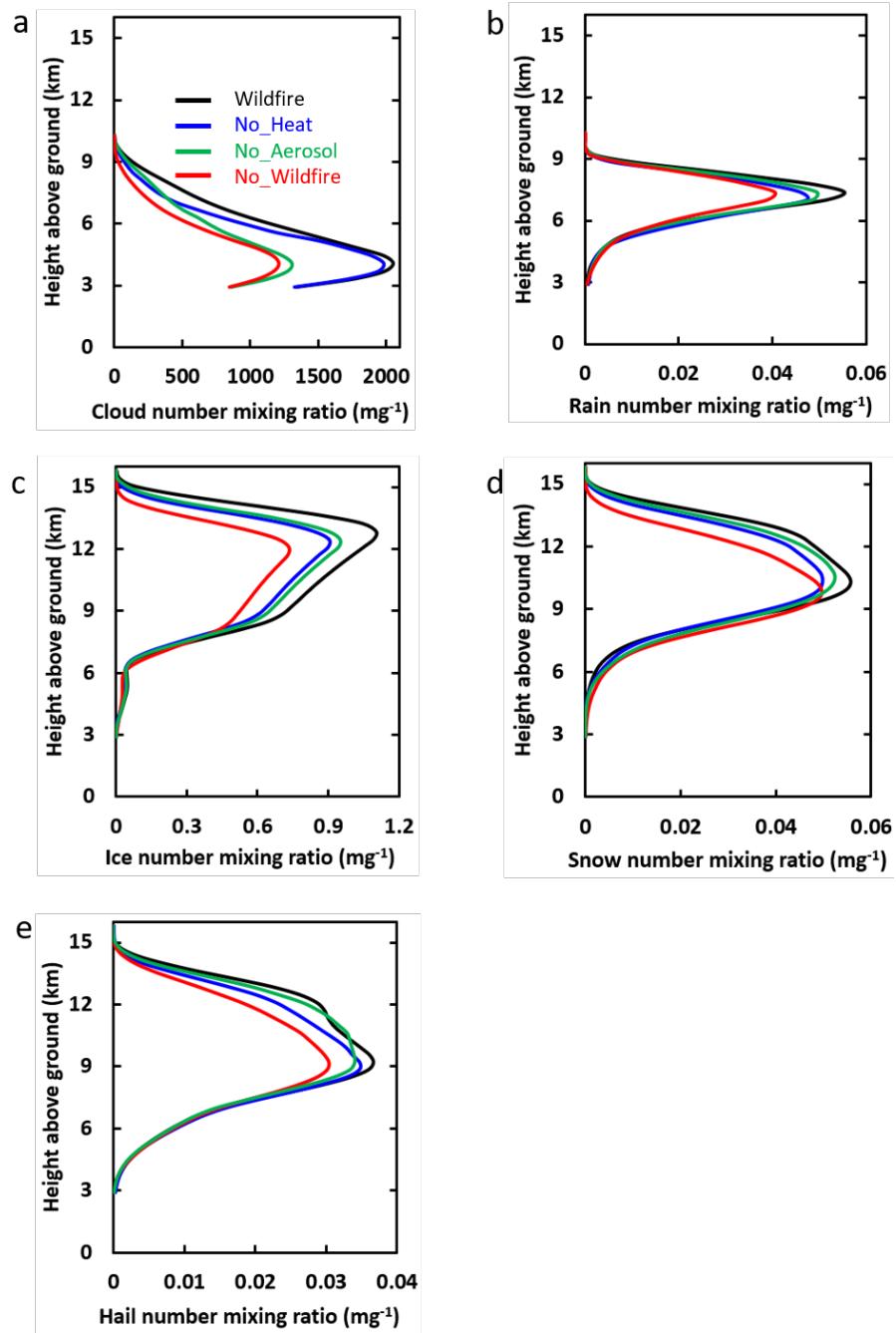
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177 **Figure S6** Time series of vertical maximum of updraft velocities (solid lines) and the
178 corresponding altitudes above ground (dashed lines) of the averaged top 25 percentile
179 updraft profiles for $w > 2 \text{ m s}^{-1}$ over the analysis domain as shown in the red box in Figure
180 S2a from the simulation Wildfire (black), No_Heat (blue), No_Aerosol (green) and
181 No_Wildfire (red).

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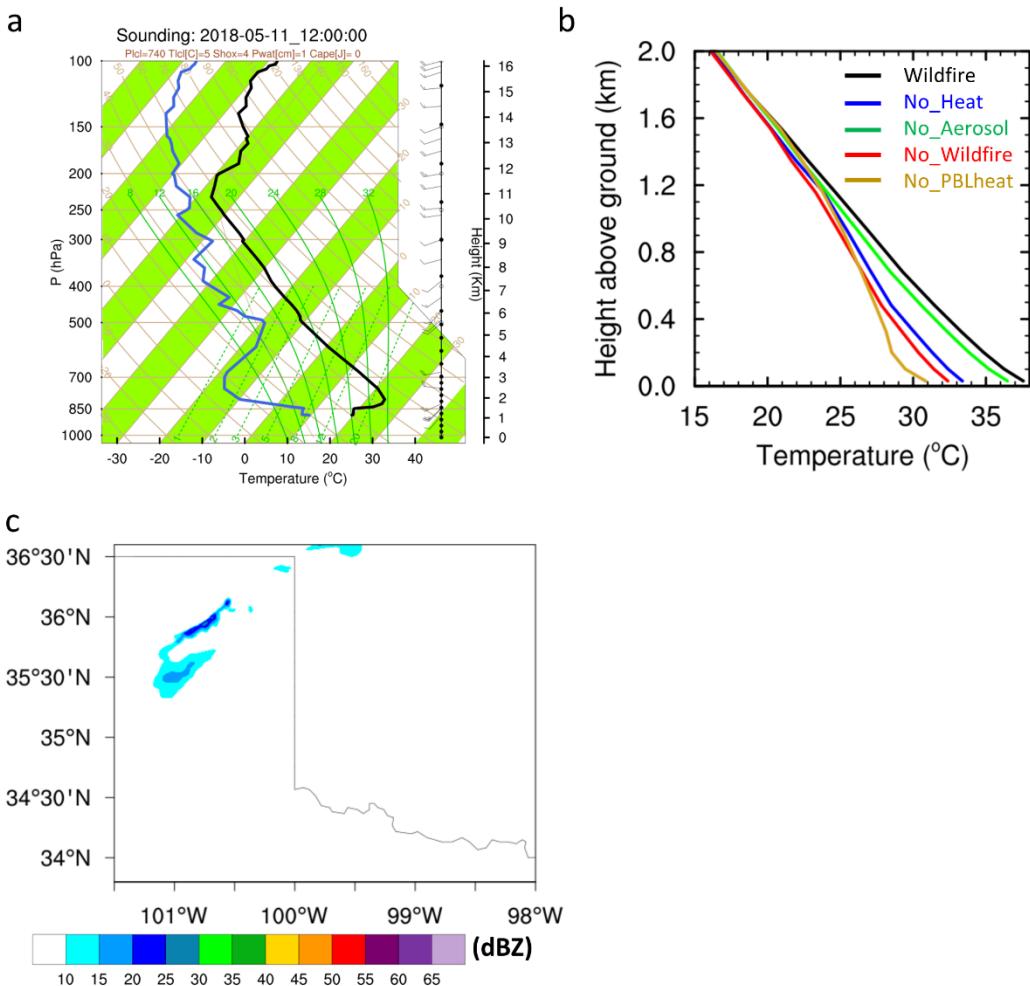


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184 **Figure S7** Vertical profiles of number mixing ratios for (a) cloud, (b) rain, (c) ice, (d) snow,
 185 and (f) hail averaged over the top 25 percentiles (i.e., 75th to 100th) of the updrafts
 186 with $w > 2 \text{ m s}^{-1}$ from the simulations of Wildfire (black), No_Heat (blue),
 187 No_Aerosol (green), and No_Wildfire (red) during the strong convection period from 2300 UTC 11
 188 May to 0000 UTC 12 May over the analysis domain as shown in the red box in Figure S2a.

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192 **Figure S8** (a) The skew-T plot for the sounding at KAMA at 1200 UTC 11 May (~8 hours
 193 before the initiation of convection). (b) Temperature profiles of Wildfire (black), No_Heat
 194 (blue), No_Aerosol (green), No_Wildfire (red), and No_PBLheat (gold) at 1800 UTC (2-
 195 hour before the initiation of convection) 11 May at the Mallard fire region (yellow box in
 196 Figure S2a). (c) Composite reflectivity from No_PBLheat at 2330 UTC when the
 197 maximum reflectivity is reached.

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