

1 **Explicit nested-grid prediction of convective-scale motions in a**
2 **skillful global model**

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8 **Key Points:**

- 9 • A global-to-regional refined atmosphere model is presented for simultaneous, skillful
10 global and convective-scale predictions.
11 • Year-round and springtime forecasts show skill equal to or better compared to opera-
12 tional global and regional models.
13 • Models based on the Finite-Volume Cubed-Sphere Dynamical Core show great promise
14 for unifying global and regional prediction systems.

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Abstract

Global convective-scale models are currently too expensive to be operationally useful, but limited-area convective-scale models are only useful for a few days' lead time. We present a new global model able to be locally refined so as to explicitly-resolve convection over a small area of the earth. This model, fvGFS, couples the GFDL Finite-Volume Cubed-Sphere Dynamical core (FV³) to the Global Forecast System (GFS) physics and initial conditions, augmented with a six-category microphysics and a modified planetary boundary layer scheme. The goal is three-fold: skillful global predictions, explicit simulation of mesoscale- and storm-scale circulations, and useful multiple-day prediction of convective-scale events, all in the same model. Here we examine the characteristics of fvGFS on a 3-km continental United States domain nested within a 13-km global model. Retrospective forecasts from all seasons are evaluated, with a focus on severe continental convection. The nested fvGFS still has good hemispheric skill comparable to or better than the operational GFS; meanwhile, convective-scale phenomena, especially supercell thunderstorms and squall lines, are explicitly represented over the refined region. In particular, fvGFS has excellent representation of fine-scale updraft helicity fields, an important proxy for severe weather forecasting. Precipitation skill is found to be superior to operational global models and competitive with operational regional models; the 3-km domain also greatly improves upon 2-m temperature and humidity biases in the global model. We discuss further development of fvGFS and the prospects of a unified global-to-regional prediction system.

1 Introduction

Advances in numerical modeling and computer capacity in the last two decades has allowed great strides in improved prediction at global and convection-permitting or convection-allowing scales [Bauer *et al.*, 2015; Clark *et al.*, 2016, and references therein]. Recently, unified global-to-regional systems [Walters *et al.*, 2017] have been discussed as a way to combine the ability of convective-scale regional models to predict individual thunderstorms and the medium-range and longer prediction capabilities made possible by a global model, to allow extended-range explicit prediction of convective storms and to reduce the dependence upon convective parameterization. Global modeling brings the benefits of longer-range prediction and relieves the need for driving boundary conditions from a global model, but also raises many challenges. Global modeling is more expensive, since the domain size is fixed, and modeling of the general circulation is challenging owing to the great diversity in world-wide phenomena and geographical features, all of which need to be correctly simulated or parameterized for a successful forecast. Regional models have the luxury of optimizing for the particular region being simulated: In particular, the characteristics of tropical and mid-latitude convection are greatly different [Emanuel, 1989], with major implications for the microphysical and convective parameterizations. A global model must maintain stability when confronted with the steep topography of the Himalayas, the Andes, and at the edges of polar ice sheets, and must also represent the powerful stratospheric winds with speeds a significant fraction of the speed of sound, violating the assumptions commonly made in regional models. A global model capable of effective simulation of convective-scale features must address all of these concerns; the need for skillful prediction only magnifies these challenges.

Our research group at GFDL has developed a new global weather forecast model, the Finite-Volume Global Forecast System (fvGFS), which replaces the spectral dynamical core used in the operational Global Forecast System (GFS) with the GFDL Finite-Volume Cubed-Sphere Dynamical core, FV³ [or FV3: Putman and Lin, 2007], in addition to other improvements to the GFS physical parameterization suite. A series of companion papers will describe this model in more detail and the many applications being developed with this model, including hydrostatic-scale prediction, tropical cyclone prediction, and global cloud-resolving modeling. This paper will demonstrate the utility of the variable-resolution capabilities of FV³, especially grid nesting, to permit the explicit simulation of convective motions over the Contigu-

66 ous United States (CONUS) in a global model; a second paper will describe results from a
67 complimentary grid-refinement strategy using a single variable-resolution grid.

68 We will show objectively that the global skill of the base global model is maintained when
69 the grid is locally refined. We will also show that fvGFS can simulate realistic mesoscale and
70 storm-scale features, including severe convection, and can predict such features both on the
71 0–36 hour timescales for which convective-scale models are currently used and also on medium-
72 range timescales of several days. The nested fvGFS described here serves as an initial pro-
73 totype of a unified global-to-regional model capable of skillful global-scale prediction and ef-
74 fective convective-scale prediction—all in the same modeling system, in the same forecast. Fur-
75 ther development will entail introduction of modern physics packages, especially revisions to
76 the microphysics and planetary boundary layer scheme, and improvements to the initializa-
77 tion, including radar and satellite data assimilation.

78 The manuscript is organized as follows. Section 2 describes the fvGFS model and its
79 configuration for the simulations described in the rest of the manuscript. Section 3 describes
80 the quantitative evaluation of the model, both global and regional forecast skill and biases, and
81 the aggregate characteristics of the simulations. Section 4 describes individual nested forecasts
82 to demonstrate the qualitative characteristics useful for prediction, especially for severe con-
83 vective events. Section 5 summarizes the paper and discusses the prospects for a unified mod-
84 eling system.

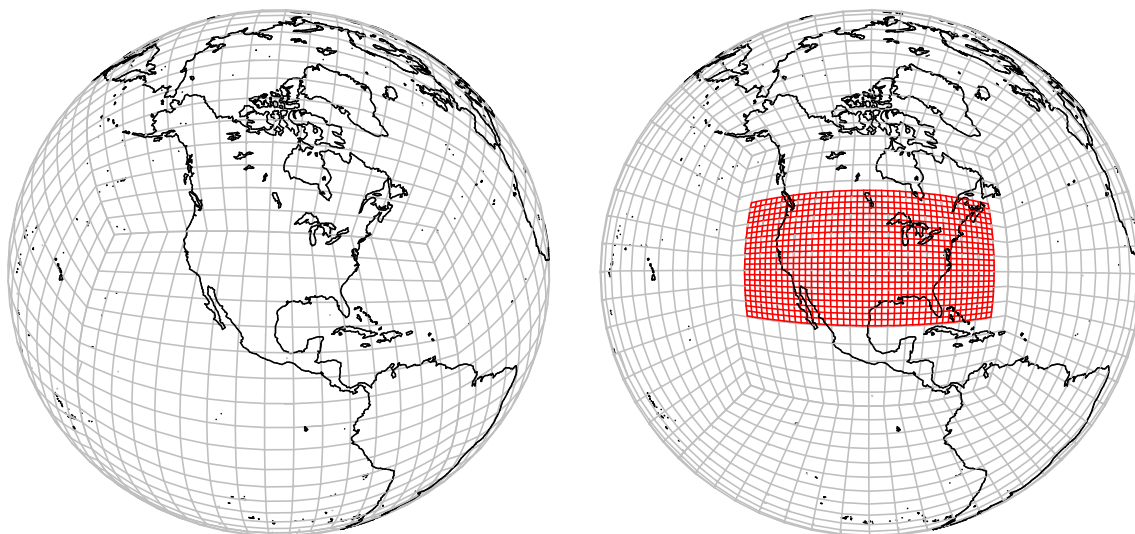
85 2 Model description and configuration

86 The fvGFS forecast model is built around the nonhydrostatic FV³ dynamical core, the
87 most recent version of FV³ and the Finite Volume (FV) core used for many years in NOAA,
88 NASA, NCAR, and elsewhere for global chemical [Bey *et al.*, 2001] and aerosol [Chin *et al.*,
89 2000] transport modeling, climate modeling [Delworth *et al.*, 2006; Zhao *et al.*, 2009; Neale
90 *et al.*, 2010; Donner *et al.*, 2011; Zhao *et al.*, 2016; Bogenschutz *et al.*, 2017], seasonal predic-
91 tion [Chen and Lin, 2011, 2013; Murakami *et al.*, 2015], real-time weather and air-quality fore-
92 casting [Suarez *et al.*, 2008], the Reanalyses MERRA and MERRA-2 [Rienecker *et al.*, 2011;
93 Gelaro *et al.*, 2017], global mesoscale and cloud-resolving modeling [Shen *et al.*, 2006; Put-
94 man and Suarez, 2011; Lin *et al.*, 2017], simulation of Martian climate [Greybush *et al.*, 2012],
95 and radiative-convective equilibrium modeling [Jeevanjee, 2017]. FV³ solves the fully-compressible
96 Euler equations using the forward-in-time scheme and Lagrangian vertical coordinate of Lin
97 [2004], the scalar advection scheme of Lin and Rood [1996] based on the piecewise-parabolic
98 method, the Lagrangian dynamics of Lin and Rood [1997], and the finite-volume pressure gra-
99 dient force of Lin [1997]. Fast vertically-propagating sound- and gravity-wave processes are
100 handled by a traditional semi-implicit solver. The horizontal discretization is on the gnomonic
101 equiangular cubed-sphere grid [Putman and Lin, 2007] capable of local refinement by two-way
102 grid nesting [Harris and Lin, 2013, 2014] and grid stretching [Harris *et al.*, 2016] through the
103 Schmidt [1977] transformation.

104 The GFS physics used for the simulations described here use the most recent scale-aware
105 Simplified Arakawa-Schubert [SA-SAS, Han *et al.*, 2017] shallow and deep convection schemes;
106 the GFS planetary boundary layer scheme [Han and Pan, 2011]; and the Rapid-Radiative Trans-
107 fer Model [RRTM, Clough *et al.*, 2005]. The simple microphysics scheme of Zhao and Carr
108 [1997] and cloud fraction scheme of Xu and Randall [1996] is replaced by the GFDL six-category
109 microphysics Chen and Lin [2013] augmented with the ability to perform fast phase changes
110 and latent heating at a faster timescale than the rest of the physical parameterizations (to be
111 described in a forthcoming manuscript by L. Zhou *et al.*). The Noah land surface model [Ek
112 *et al.*, 2003], upgraded to use high-resolution land surface data [Wei *et al.*, 2017], provides land-
113 surface interactions.

114 The nested-grid configuration of fvGFS used here uses a global cubed-sphere grid with
115 768 by 768 grid cells (c768) on each of the six sides of the cube, rotated and stretched by a

116 factor of 1.5 to reach a grid-cell-width of roughly 9 km over the CONUS. A factor-of-three
 117 two-way nest is then placed over the CONUS, yielding 3-km grid-cell width (Figure 1). The
 118 model uses 63 Lagrangian vertical levels on both grids, each using the GFS hybrid-pressure
 119 levels as the reference vertical coordinate; the first model layer is centered between 15 and 25
 120 m above the surface, depending on the atmospheric conditions, with 14 layers below 1.5 km
 121 and 19 below 3 km. On the coarse global grid, the tracer advection, vertical remapping, and
 122 physical parameterizations use a 90-s timestep, with an acoustic-mode timestep of approxi-
 123 mately 12.5 s. On the nested grid, the same 90-s physics timestep is used, but the tracer ad-
 124 vection, vertical remapping, and fast microphysical processes use a 22.5-s timestep, and the
 125 acoustic-mode timestep is 4.5 s. This choice of timesteps is stable for all of the simulations
 126 described in this paper, while remaining efficient and also allowing us to perform the fast la-
 127 tent heating processes at a frequency appropriate for convective scales. Interactions between
 128 the nested and coarse grids (boundary conditions and two-way updating) occur every 90 s. The
 129 radiation scheme is updated hourly.



130 **Figure 1.** Structure of uniform global grid (left) and global stretched cubed-sphere (gray) and CONUS
 131 nested (red) grids (right).

132 Some modification of the GFS physics on the nested grid is necessary for best results
 133 at convective scales. We disable the convective parameterization (both shallow and deep) on
 134 the nested grid, unless noted otherwise. To address issues with the GFS PBL scheme on con-
 135 vective scales [Hong *et al.*, 2006], we constrain the local vertical mixing in the boundary layer
 136 by reducing the parameterized turbulent diffusivity by half, and disable the diffusion in the in-
 137 version layer. Further improvements to the boundary layer scheme are being developed.

138 The model is currently cold-started from operational GFS analyses, of about 13 km nom-
 139 inal resolution; no data assimilation or other initialization techniques are used.

140 We present model forecasts for two time periods. One is a 74-case set of “hindcasts”
 141 initialized at 00Z every five days from 16 January 2015 to 15 January 2016. This set of fore-
 142 casts include a very broad range of synoptic-scale regimes across all seasons, and permits a
 143 very robust evaluation of forecast skill and model robustness. We have found that hemispheric
 144 forecast skill can vary significantly on a month-to-month basis, and skill in a 30- or even 60-
 145 day period may be unrepresentative of the year-round skill. The simulations are 7 days long
 146 during this time period, and use two configurations, one with (“SA-SAS nested fvGFS”) and

147 one without (“nested (noconv) fvGFS”) SA-SAS enabled on the nested grid; otherwise the two
 148 configurations are identical. (Unless otherwise stated all nested model results refer to nested
 149 noconv fvGFS.)

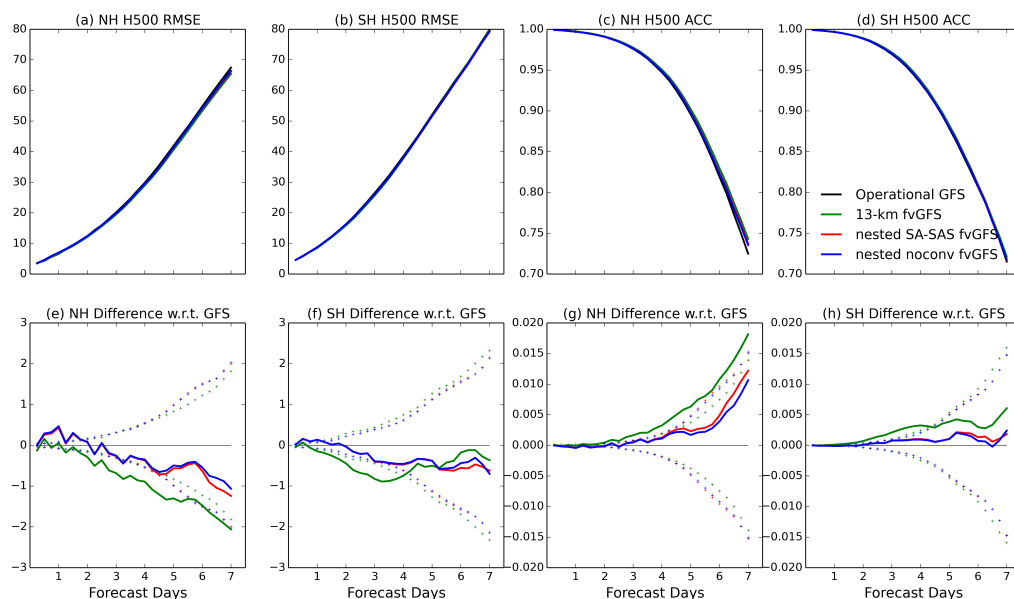
150 The second time period covers the 77 forecasts run daily at 00Z from 01 April to 16 June
 151 2017 for the Hazardous Weather Testbed (HWT) Spring Forecast Experiment (henceforth sim-
 152 ply “Spring Experiment” period), which imposes a requirement to have forecasts finished, post-
 153 processed, and delivered to HWT by roughly 7am CDT each day. Only the nested noconv con-
 154 figuration was run for this period. This time period included a number of significant severe-
 155 weather outbreaks, which are useful cases to show the capabilities of fvGFS. All nested fvGFS
 156 simulations during this period are integrated to 120 hours.

157 **3 Forecast skill and aggregate simulation characteristics**

158 The coarsest-grain metric for forecast skill is that of the hemisphere-wide 500-hPa height.
 159 A good 500-hPa skill does not necessarily mean that a given model will give better forecasts
 160 of impactful weather, but poor 500-hPa skill indicates degraded forecasts, especially at longer
 161 lead times. Plots of Northern and Southern Hemisphere root-mean square error (RMSE; K)
 162 and anomaly correlation coefficient [ACC; *Murphy and Epstein, 1989*] for the 2015 all-season
 163 hindcasts are shown in Figure 2, for the two sets of nested hindcasts. For reference, they are
 164 compared to both the operational GFS as well as the GFDL real-time 13-km uniform-resolution
 165 fvGFS (henceforth referred to as the 13-km fvGFS), which has been aggressively optimized
 166 for large-scale skill. Table S1 summarizes the configurations of all of the models mentioned
 167 in this paper. For this time period, the 13-km fvGFS has RMSE and ACC which are statistically-
 168 significantly improved over the operational GFS (lower RMSE and higher ACC), especially
 169 after day 5. The two nested hindcast suites, which have not been as optimized for global sim-
 170 ulation, show a slightly poorer skill but still better than the GFS after the first two days. This
 171 reduction in skill is likely due to the mismatch in resolution between the convective scales over
 172 the CONUS needing to spin up from the coarser GFS analyses; overall, there is a minor degra-
 173 dation of the hemispheric 500-hPa skill in a two-way nested run. The runs during 2017 Spring
 174 Experiment period (Figure S1) have a more modest improvement over the operational GFS,
 175 owing to improvements in the operational model and month-to-month forecast skill variabil-
 176 ity.

182 Of more direct importance to human activity are surface measures of temperature and
 183 humidity. RMSE and bias for 2-m temperature (T2m) and 2-m dew point temperature (DPT2m)
 184 over the eastern and western US are shown in Figure 3 for the 2015 hindcasts; Figures 4 and
 185 5 show the spatial distribution of the biases, for a 24-hour forecast and averaged over the en-
 186 tire forecast period, respectively. The CONUS T2m errors are significantly lower in the nested
 187 fvGFS than in the operational GFS, while the 13-km fvGFS shows smaller but still signifi-
 188 cant improvements over the GFS. The nested fvGFS shows a significant reduction in cold and
 189 dry biases compared to the operational GFS; the 13-km fvGFS shows a smaller improvement
 190 to these biases. In the western US the nested fvGFS shows a significant reduction in cold bi-
 191 ases, especially over the intermountain region and Pacific Northwest, although a nighttime warm
 192 bias has emerged in the Great Plains region. The errors and biases in nested fvGFS were nearly
 193 identical in the SA-SAS nested and nested noconv configurations, demonstrating the scale-awareness
 194 of SA-SAS.

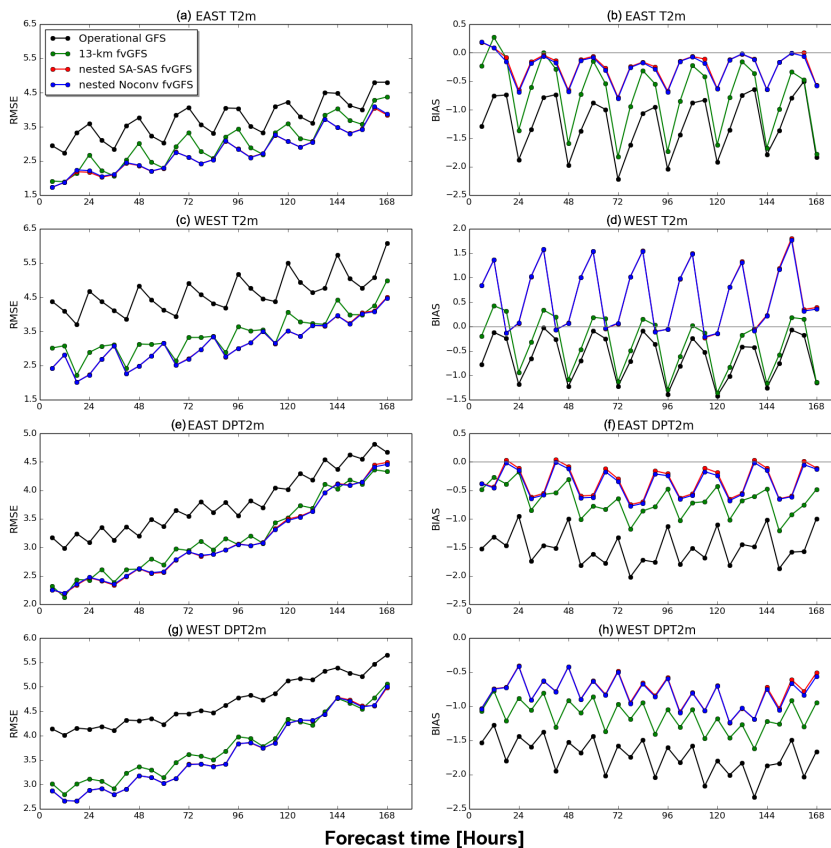
203 The most important and most difficult to forecast impact is precipitation. Figure 6 shows
 204 the skill scores, measured by equitable threat score [ETS; *Wang, 2014*], fractions skill score
 205 [FSS; *Roberts and Lean, 2008*], and bias score compared to observed precipitation from the
 206 Stage IV multi-sensor quantitative precipitation estimate [*Lin and Mitchell, 2005; Lin, 2011*]
 207 for both sets of 2015 year-round hindcasts; here, we present both ETS and FSS to show that
 208 our results are not specific to a certain measure of precipitation skill. As expected, we find that
 209 both ETS and FSS are highest at about 12–24 hours after initialization and for the lightest pre-
 210 cipitation rates, and then decreases steadily for longer lead times and higher rates. ETS and



177 **Figure 2.** Hemisphere-wide 500-hPa height skill for 2015 year-round hindcasts. Top row: (a) northern-
 178 hemisphere (NH: 20–80 N latitude) and (b) southern-hemisphere (SH: 80–20 S latitude) root-mean square
 179 error, lower is better; (c) northern-hemisphere and (d) southern-hemisphere anomaly correlation coefficient,
 180 higher is better. Bottom row: (e–h) as in (a–d) but relative to the contemporary operational GFS; dotted lines
 181 represent 95% confidence interval for the differences from GFS.

211 FSS are highest for the period 18Z-00Z each day, and is slightly higher in the SA-SAS nested
 212 fvGFS. For both versions of the nested fvGFS, the bias scores are close to 1 (optimal) for light
 213 and moderate precipitation, with a low bias for heavy precipitation rates (over 25 mm/6hr; Fig-
 214 ure 4f); in all cases the nested fvGFS has smaller biases than the 13-km models. The SA-SAS
 215 nested fvGFS shows modestly reduced low bias for heavy precipitation but also more light pre-
 216 cipitation; the skill scores are nearly the same. The nested fvGFS substantially improves upon
 217 the biases of the 13-km fvGFS for all precipitation thresholds but has lower skill, especially
 218 in FSS and for lighter precipitation rates. All of the fvGFS versions show a substantial im-
 219 provement over the operational GFS on the wet bias over the western US (Figure 7), although
 220 the improvement of the nested model over the 13-km fvGFS is more minor except over the
 221 southwestern US.

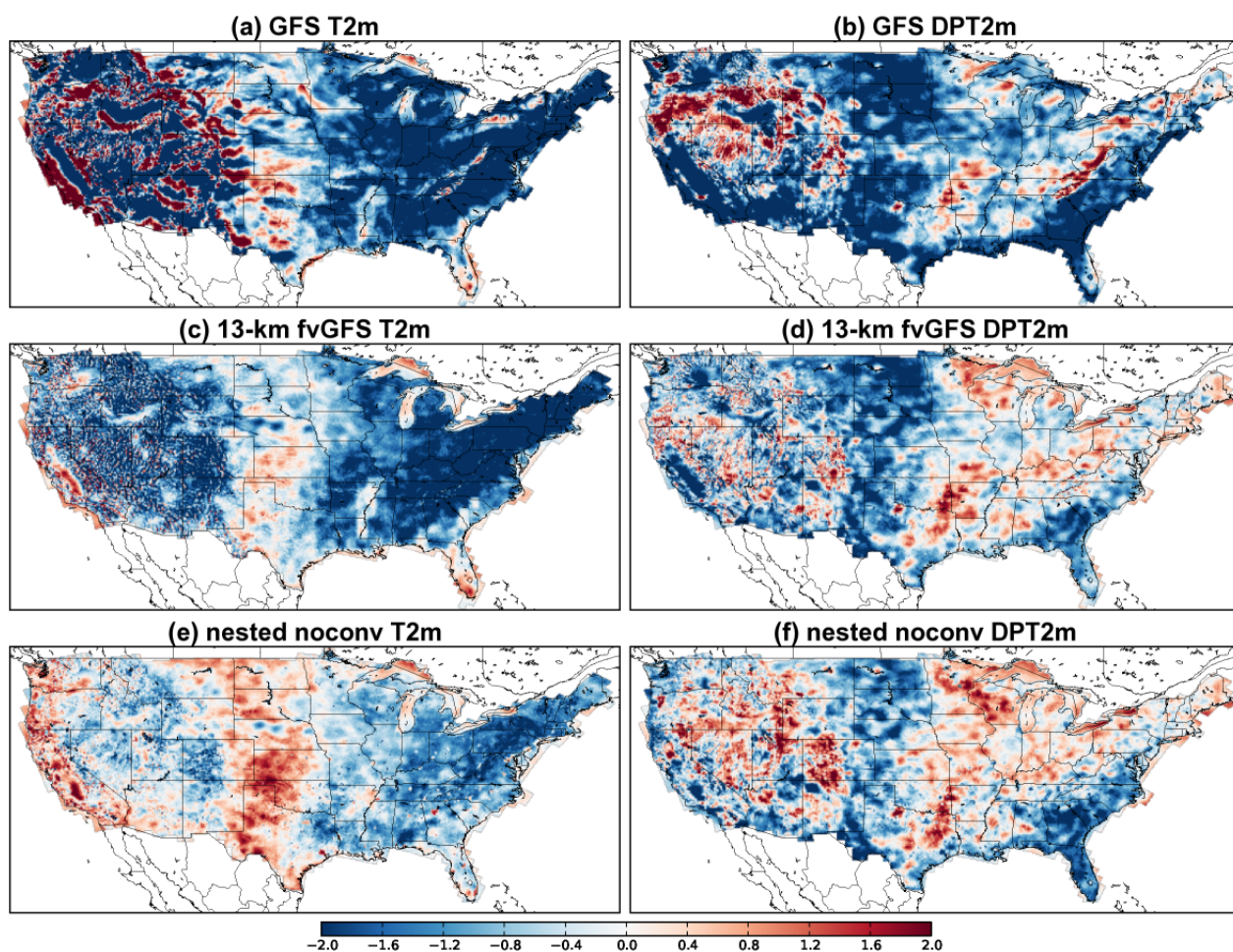
228 We can also compare fvGFS to the operational 3-km nest of the North American Model
 229 (NAM; see overview at http://www.emc.ncep.noaa.gov/mmb/mmbp11/misc/NAM_2017.pdf)
 230 during the Spring Experiment period (2015 NAM output was no longer available as of this
 231 writing). The skill and biases of the nested fvGFS are largely comparable to those of the NAM
 232 (Figure 8); in fact, both fvGFS versions have slightly better skill for light precipitation rates,
 233 and the NAM has a high bias in heavy precipitation while nested fvGFS has a low bias. The
 234 nested fvGFS has similar biases to the NAM in the eastern two-thirds of the US, and reduced
 235 high biases in the Western US, as well as less noise in regions of complex terrain (Figure 9).
 236 We can also compare the temperature and dew point errors to those of the NAM during the
 237 Spring Experiment period (Figures S2 and S3); at 24 hours the temperature errors over the East-
 238 ern US are comparable to those of the nested fvGFS, but the NAM has lower warm and dry
 239 biases over the central plains.



195 **Figure 3.** RMSE (K; a,c,e,g) and bias (K; b,d,f,h) for Eastern US 2-m temperature (a,b), Western US
 196 2-m temperature (c,d), Eastern US 2-m dew point temperature (e,f), and Western US 2-m dew point tem-
 197 perature (g,h) during the 2015 year-round hindcasts. The definitions of east and west regions are given at
 198 <http://www.emc.ncep.noaa.gov/mmb/research/nearsfc/nearsfc.verf.html> .

244 A broader indication of the behavior of individual systems can be seen in the time-longitude
 245 plots in Figure 10, depicting a meridionally-averaged plot of precipitation from each day’s nested
 246 fvGFS forecasts at different lead times during an active period in May 2017. Both rapidly-moving
 247 thunderstorm outbreaks and the more slowly-propagating but longer-lived synoptic-scale sys-
 248 tems are apparent, and compare very well with Stage IV in timing, positioning, and strength.
 249 The consistency between the forecasts at different lead times is remarkable, and numerous out-
 250 breaks are correctly predicted even 84–108 hours in advance of the event. When compared
 251 against the 13-km fvGFS (Figure S4), it is apparent that although both models are able to cap-
 252 ture the broad areas of precipitation, the nested fvGFS has a much better representation of small
 253 propagating and more rapidly-propagating systems, which appear as streaks with a shallower
 254 angle in Figure 10. This difference between the parameterized and explicit convection is most
 255 noticeable for the event on 20–21 May, which even four days in advance appears as a coher-
 256 ent, rapidly-propagating feature in the nested fvGFS.

261 We can average a number of forecasts to determine whether fvGFS captures the diurnal
 262 cycle of precipitation. Figure 11a,b show the meridionally-averaged precipitation averaged
 263 over each forecast during May 2017 as a function of forecast hour for the 13-km (a) and the
 264 nested (b) fvGFS, compared to the observed diurnal cycle (Figure 11c) during the same month.
 265 The broad-scale features, particularly the eastward propagation of systems, the evening (03z
 266 to 06z) maximum in precipitation over the Great Plains (100W to 90W), and the morning-to-
 267 afternoon precipitation over the Eastern US, are all captured by the nested model; the 13-km



199 **Figure 4.** Biases of 2-m temperature (T2m, K; a,c,e) and 2-m dew point temperature (DPT2m, K; b,d,f) at
 200 24 hours after initialization during the 2015 year-round hindcasts for the operational GFS (a,b), 13-km fvGFS
 201 (c,d), and nested noconv fvGFS (e,f).

268 model's parameterized convection shows some sign of eastward propagation, albeit less co-
 269 herent and typically weaker. The amounts of precipitation are largely faithfully reproduced by
 270 the model, except for a low bias between 85W and 75W. The amplitude of the simulated di-
 271 urnal cycle in the nested model is less than observed over the western half of the region (90
 272 to 110 west longitude; Figure 11d) and greater than observed in the eastern half (70 to 90 west
 273 longitude; Figure 11e), although the 13-km model shows very little diurnal amplitude in the
 274 eastern half.

281 **4 Mesoscale and storm-scale model characteristics of nested fvGFS**

282 In the previous section the objective forecast skill of the nested fvGFS has been demon-
 283 strated. We now show specific examples of the characteristics of predicted weather systems
 284 in the 3-km nested fvGFS. We will focus on warm-season severe weather since the 3-km runs
 285 were specifically designed for the Spring Experiment. Unless otherwise noted, all references
 286 to the model are to nested (noconv) fvGFS

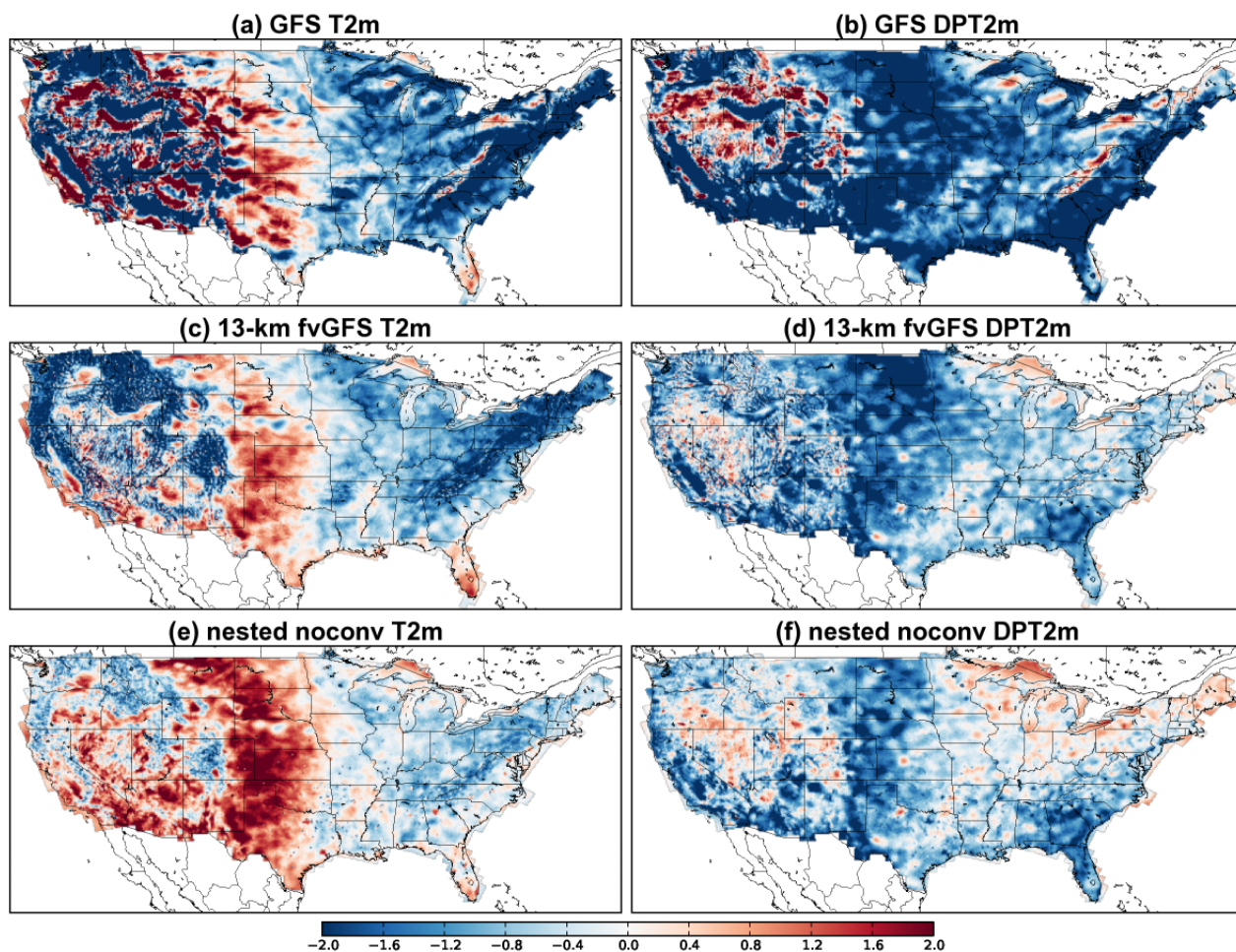


Figure 5. As in Figure 4, but for the entire 0–5 day forecast period.

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4.1 Multi-scale predictability: 30 April–1 May 17 squall line

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A frontal squall line is an example of an event which is strongly forced by synoptic-scale disturbances but also embeds many small-scale cells, which can alter the propagation of its neighboring storms and can feed back onto the mesoscale organization of the line. Figure 12 depicts an event on 30 April–1 May 2017 in which a long-lived cold-frontal squall line is propagating eastward from west of the Mississippi river through the Gulf Coast states. The fvGFS forecast correctly predicts the positioning of the main squall line at 18Z and 00Z (Figure 12a,b) while correctly depicting the long line of individual convective cells at the front of the squall line, stretching from Tennessee down to the Gulf coast, with a substantial trailing stratiform region of lighter reflectivities, and correctly predicts the decay of the northern part of the line by 00Z. The development of a post-frontal squall line by 06Z is also predicted, although the model generates it much earlier than observed and strengthens it much more slowly, with intense reflectivities not seen until after 06Z.

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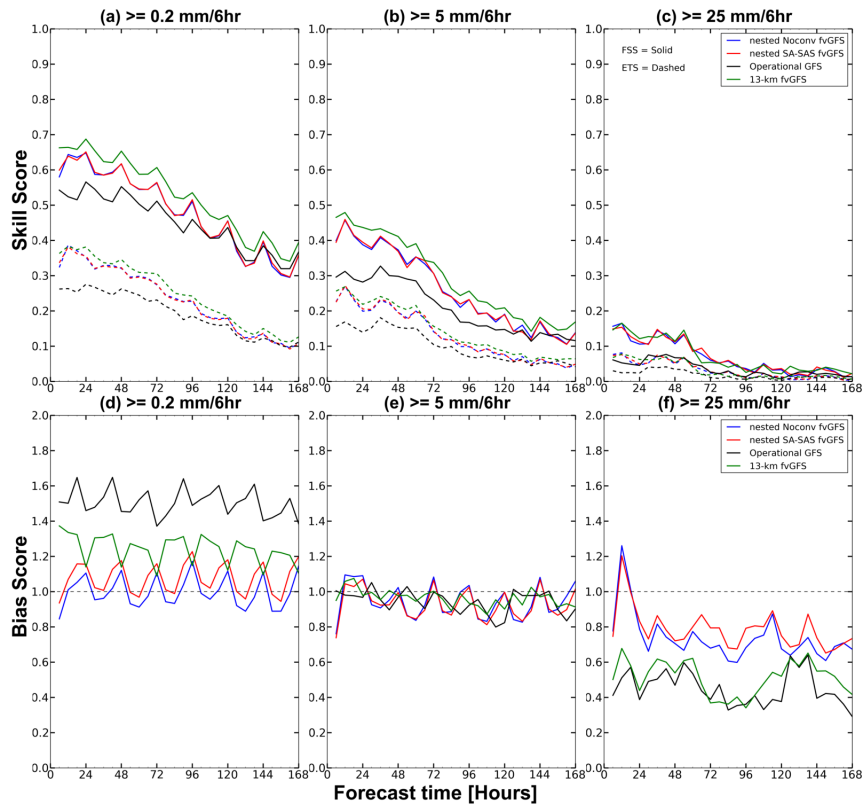
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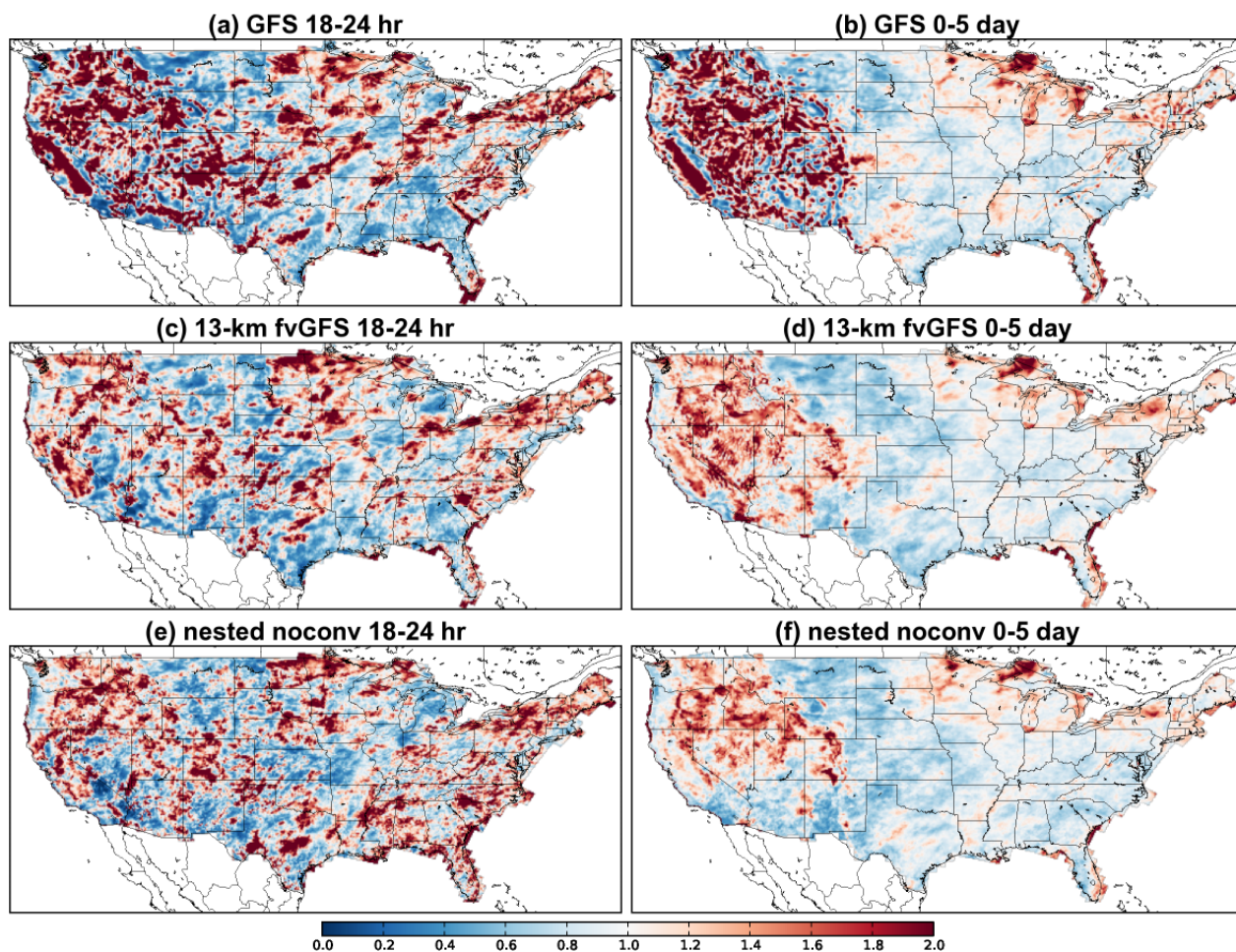
Since the model is cold-started from a coarse-resolution GFS analysis, a spin-up period is necessary before storm-scale features are fully developed. However it is interesting to examine the behavior of the model as it equilibrates. Figure 13 shows the same 1 May event but for initialization 00Z on 1 May. In the first hour after initialization (Figure 13a), the structure of the squall line is very similar to what we might expect from a hydrostatic-scale model, although some small-scale convection in unstable areas is already beginning to appear in Mis-



222 **Figure 6.** Comparison of Skill scores (a–c; ETS dashed, FSS solid), and Bias scores (d–f) between the two
 223 nested fvGFS configurations and the operational GFS for the 2015 year-round hindcasts, for precipitation
 224 rates greater than or equal to 0.2 (a,d), 5.0 (b,e) and 25.0 (c,f) mm/6 hr. All models are re-gridded to 4-km
 225 resolution to match the Stage IV verification and a 12-km neighborhood is employed for ETS and FSS.

309 mississippi and in Georgia. The GFS analysis contains no precipitating hydrometeors and thus would
 310 be devoid of radar echoes; so all of the precipitating hydrometeors are produced by fvGFS dur-
 311 ing its equilibration. Later times (Figure 13b–d) show more small-scale detail as well as bet-
 312 ter organization of mesoscale features, such as the multiple contiguous bands of moderately-
 313 strong echoes, lines of discrete cells, and the formation of stratiform regions. Fortunately, the
 314 placement and structure of some of these features is very accurate, especially at 09Z (Figure 13c).

317 A more rigorous examination of model spin-up can be made by computing the kinetic
 318 energy spectrum [cf. *Koshyk et al.*, 1999] of the nested-grid region. Figure 14 shows the spec-
 319 trum of 250 mb kinetic energy computed using the two-dimensional discrete fourier transform
 320 of *Denis et al.* [2002], which at these resolutions should roughly follow a $-5/3$ slope [*Nastrom*
 321 *and Gage*, 1985]. At one hour after initialization (blue) the larger scales (> 100 km) present
 322 in the analysis are represented in the model, but there is much less activity at smaller scales
 323 than we would expect from a $-5/3$ spectrum. The conspicuous exception is near the $4\Delta x$ cut-
 324 off of representable scales by the model, at which some small-scale motions (presumably con-
 325 vection, as in Figure 12a) has already started. By three hours after initialization (green), the
 326 intermediate (20–100 km) scales have largely been filled-in, as have the marginally-resolved
 327 scales ($4\Delta x$ – $6\Delta x$), with a sharp cutoff of the poorly-resolved scales below $4\Delta x$; by six hours
 328 (red), the spectrum now closely resembles the average (heavy black line) over the rest of the
 329 five-day forecast. A similar spin-up of Hurricane Harvey (2017) within a few hours has also
 330 been seen in a different 3-km nested configuration (A. Hazelton, personal communication). Al-
 331 though this rapid spin-up may not occur for all regimes, particularly more weakly-forced warm-



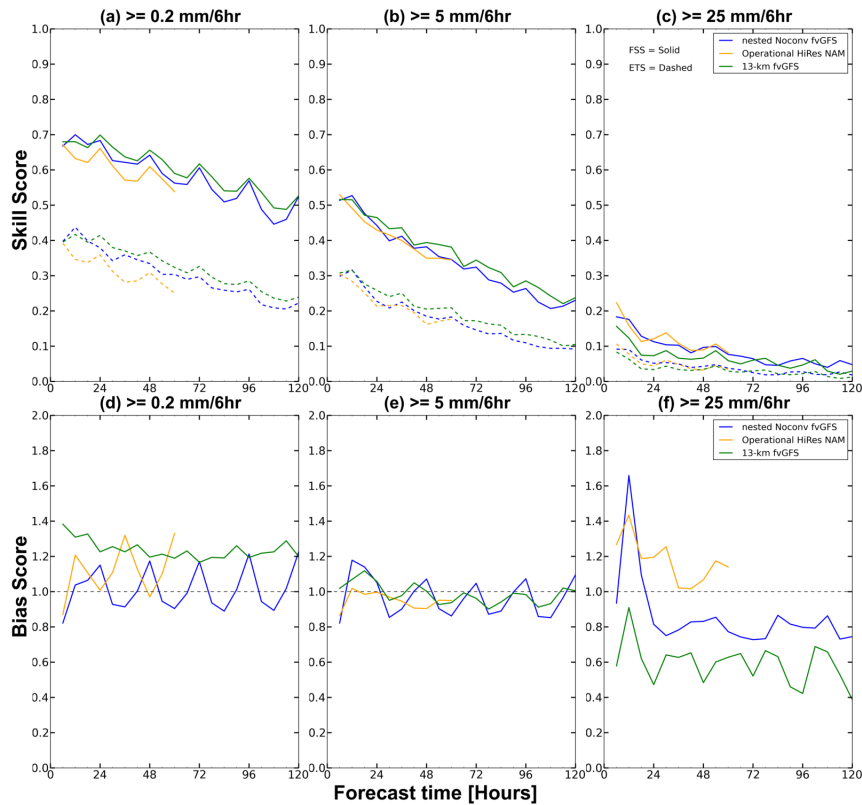
226 **Figure 7.** As in Figure Figure 4, but showing ratio of modeled to observed precipitation (fractional bias; 1
 227 implies no bias) during the 2015 year-round hindcast period.

332 season events, this result does show the possibility for very rapid equilibration in fvGFS, use-
 333 ful for both data assimilation and nowcast applications.

334 There is still substantial variability in the magnitude and the shape of the spectrum even
 335 during this five-day forecast, especially at the smallest resolved scales; further experiments (Fig-
 336 ure 14, right) representing all seasons show that there can be strong regime-dependent shifts
 337 in the spectral shape, spectral slope, and even total energy of the nested region, which are ex-
 338 acerbated if the region on which the analysis is performed is reduced. This is the opposite of
 339 our experience with globally-uniform fvGFS simulations [*Lin et al.*, 2017] which typically have
 340 very robust global kinetic energy spectra across runs, even at uniform global 3-km resolution.
 341 These results suggest that caution is warranted in interpreting spectra from limited-area mod-
 342 els for limited numbers of events or over a small domain.

347 **4.2 Storm-scale structure: 18 May 17 Central Plains supercells and squall line**

348 A compelling advantage of convective-scale models is that they are able to explicitly simu-
 349 late the convective mode, providing useful information as to the likelihood of particular haz-
 350 ards. While the predictability of any one cell is very low, this sort of model can be used to
 351 predict the region in which severe thunderstorms may occur. However, while 3-km grid-spacing



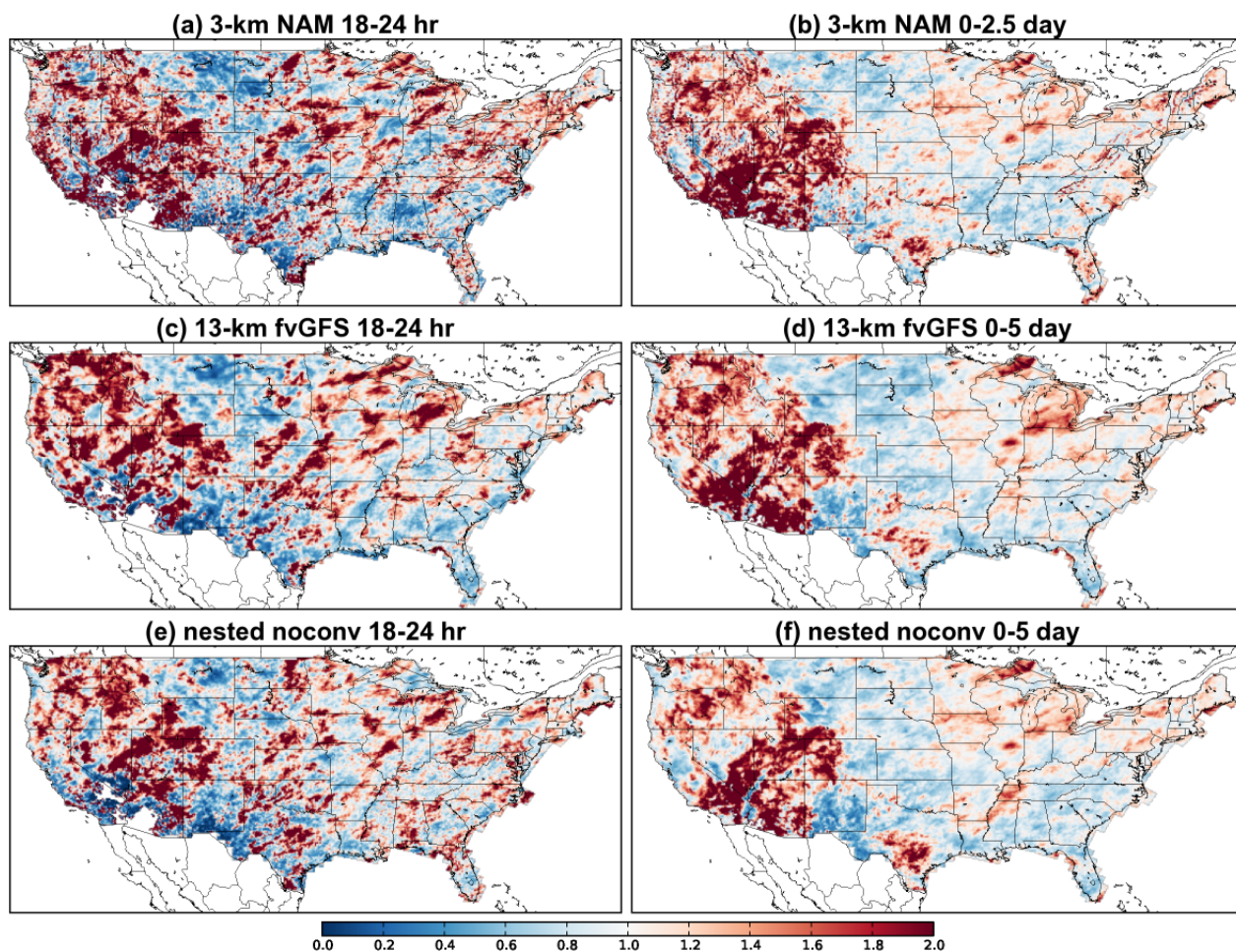
240 **Figure 8.** As in Figure 6 but for the 2017 Spring Experiment period and comparing the 13-km and the
 241 nested fvGFS to the operational NAM.

352 is able to resolve rotating updrafts, it cannot resolve tornadoes or microbursts, requiring the
 353 use of proxy diagnostics such as updraft helicity to determine the possibility of hazardous events.

354 Figure 15 shows the predicted and observed composite reflectivities for a supercell out-
 355 break in western Oklahoma into southern Kansas on the evening of 18 May 2017. The nested
 356 fvGFS correctly predicts the initiation of discrete cells along the Oklahoma-Texas border at
 357 20Z, and correctly predicts the evolution into an elongated squall line after 00Z with a cyclonic
 358 turning over Nebraska; the line does propagate more quickly than observed, with too-early ini-
 359 tiation over Texas.

361 Figure 16 shows a close-up of storm-scale structure in the Oklahoma supercells shortly
 362 after initiation. Several individual cells are visible from the composite (Figure 16a) and base
 363 (Figure 16b) reflectivity fields, with broad stratiform regions of moderate (20–40 dBz) and vis-
 364 ible convective cores, although the base reflectivities of the cores are weaker than observed.
 365 (A forthcoming manuscript will describe a similar set of nested fvGFS simulations using the
 366 microphysical scheme of *Thompson et al.* [2008], in which the convective cores are much stronger
 367 but the stratiform regions less extensive). Both the cores and the stratiform regions are vis-
 368 ible in the column-integrated condensate (Figure 16c), indicating large amounts of water and
 369 ice near the core and smaller amounts into the stratiform regions.

373 The coherent structures in the hourly-maximum of composite reflectivity (Figure 16d)
 374 indicates that intense, marginally-resolved moving or propagating systems are being correctly
 375 represented in fvGFS. To see this, we examine hourly-maximum updraft helicity (UH), the ver-
 376 tical integral of the product of vertical velocity and vertical vorticity (Figure 17a–d), and the
 377 hourly-maximum column-maximum updraft (Figure 17e–h). Numerous coherent, elongated



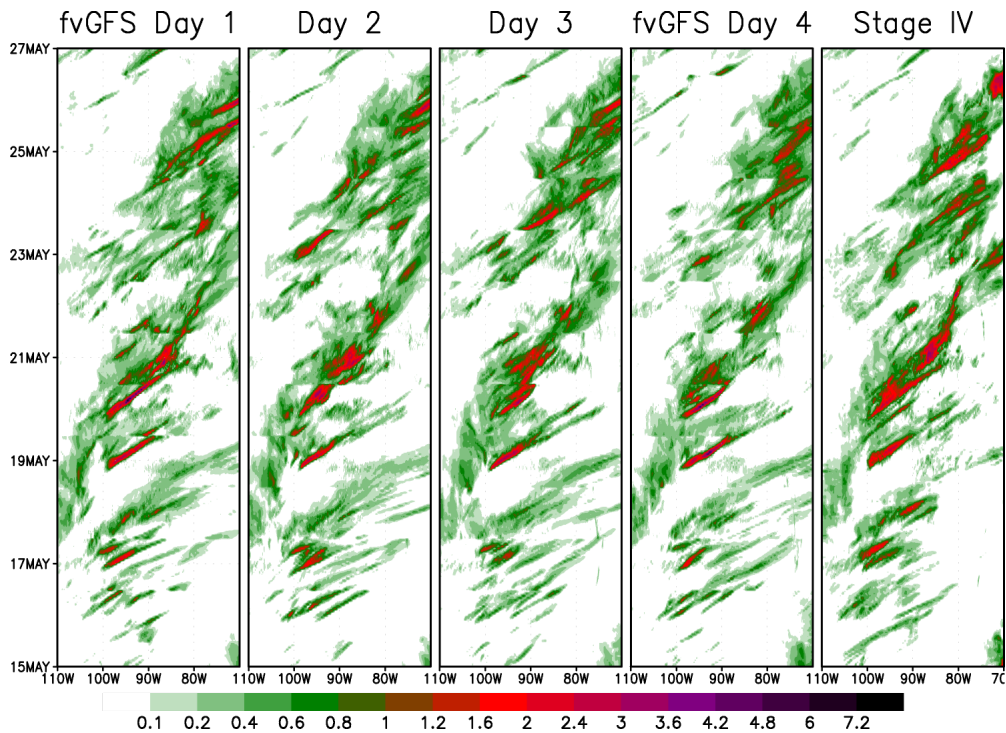
242 **Figure 9.** As in Figure 7 but for the Spring Experiment period, and comparing to the operational NAM
 243 nest, for which totals are shown for the 60 hour duration of the NAM forecast.

378 streaks are visible in the UH field, with some features only one or two grid cells wide. These
 379 features are maintained for an extended period as they move with their parent storm; they are
 380 well-represented rotating updrafts in the model, and not numerical noise, despite their very small
 381 size. The same finely-detailed features are visible in the hourly- and column-maximum up-
 382 drafts, indicating that both vorticity and vertical velocity are well-represented in fvGFS, even
 383 near the grid scale.

384 FV^3 was designed to accurately represent and advect the (vertical) vorticity and any of
 385 its products with a cell-average scalar as if it was itself a scalar, as described for shallow-water
 386 potential vorticity (mass times vorticity) by *Lin and Rood* [1997]; in particular fine-scale fea-
 387 tures and sharp gradients are very well preserved. Updraft helicity is the product of vorticity
 388 and vertical velocity, both cell-mean quantities in FV^3 , and so fvGFS is able to represent co-
 389 herent, long-lived features in this field even near the grid scale.

392 **4.3 Multi-day prediction: 27–28 May 17 “Triple Derecho”**

393 Derechos are intrinsically driven by intense discretely-propagating thunderstorm cells
 394 and are very difficult to represent with the parameterized convection of current medium-range
 395 prediction models. We present forecasts of the “triple derecho” of 27-28 May 2017 (Figure 18)

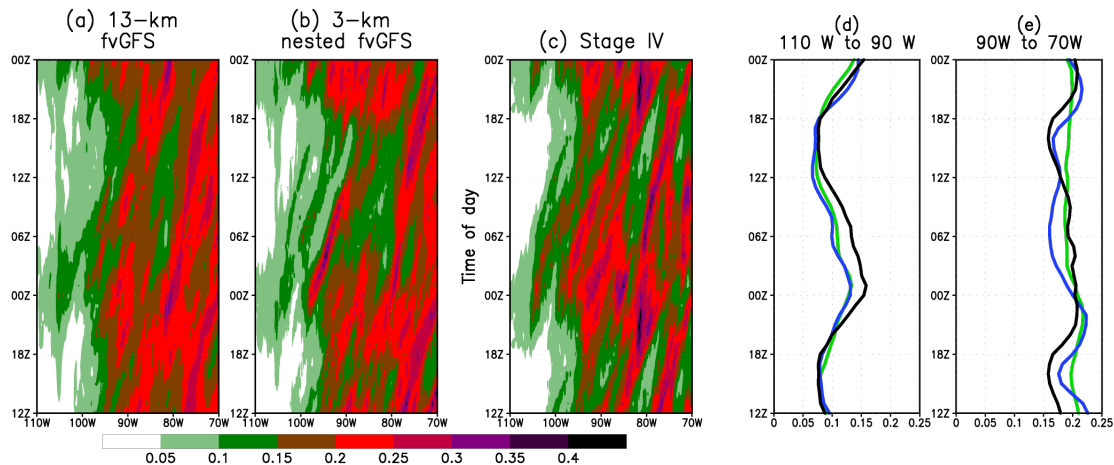


257 **Figure 10.** Time-longitude plots of hourly-accumulated precipitation (mm) in daily nested fvGFS forecasts,
 258 for different lead times of 12–36 hours (day 1), 36–60 hours (day 2), 60–84 hours (day 3), and 84–108 hours
 259 (day 4), compared to Stage IV multi-sensor observed precipitation. The results are averaged from 30 to 50
 260 latitude, for the period of 15 to 27 May 2017.

396 as first identified by the Storm Prediction Center (SPC), in which three individual events in
 397 the Ohio valley were observed, each fulfilling the *Corfidi et al.* [2016] definition of derecho.
 398 The forecast initialized at 00Z that day (Figure 19) clearly shows three extensive, rapidly-propagating
 399 bowing features; the first moving through Kentucky and Tennessee, the second moving from
 400 Missouri through Tennessee into Alabama and Georgia, and the third moving through Arkansas,
 401 all broadly in agreement with the timing and positioning of the observed events. The maxi-
 402 mum values of the 2–5 km updraft helicity during this event (Figure 20a) shows the tracks of
 403 the successive systems through this region, although the axes of maximum helicity values are
 404 displaced slightly southward of the observed storm reports.

410 This derecho event was predictable several days in advance, longer than the longest-lead
 411 forecast of operational US convection-allowing models. Predictions 48, 72, and 96 hours in
 412 advance (Figure 21) all show multiple bow echo signatures reminiscent of derechos in the Ohio
 413 and lower Mississippi Valleys, although the precise timing varies and there is a tendency to-
 414 wards too much activity in eastern Texas. The forecast initialized on 24 May (Figure 20b) again
 415 shows large regions of high updraft helicities organized along a couple of main tracks, although
 416 oriented more from the southwest to the northeast than the later forecast (Figure 20a) and with
 417 less activity in Missouri.

420 The environmental conditions to predict this derecho were predictable in both the 3-km
 421 nested and 13-km fvGFS. Figure 22 shows that a synoptic-scale environment favorable for bow
 422 echoes and derechos was predicted up to five days in advance of the event, with strong uni-
 423 directional shear and high convective available potential energy in Missouri and Arkansas and
 424 eastward through the Ohio Valley. That the environment was well-predicted is substantiated
 425 by the predictions from the real-time 13-km fvGFS shown in Figure 23: a squall line is cor-



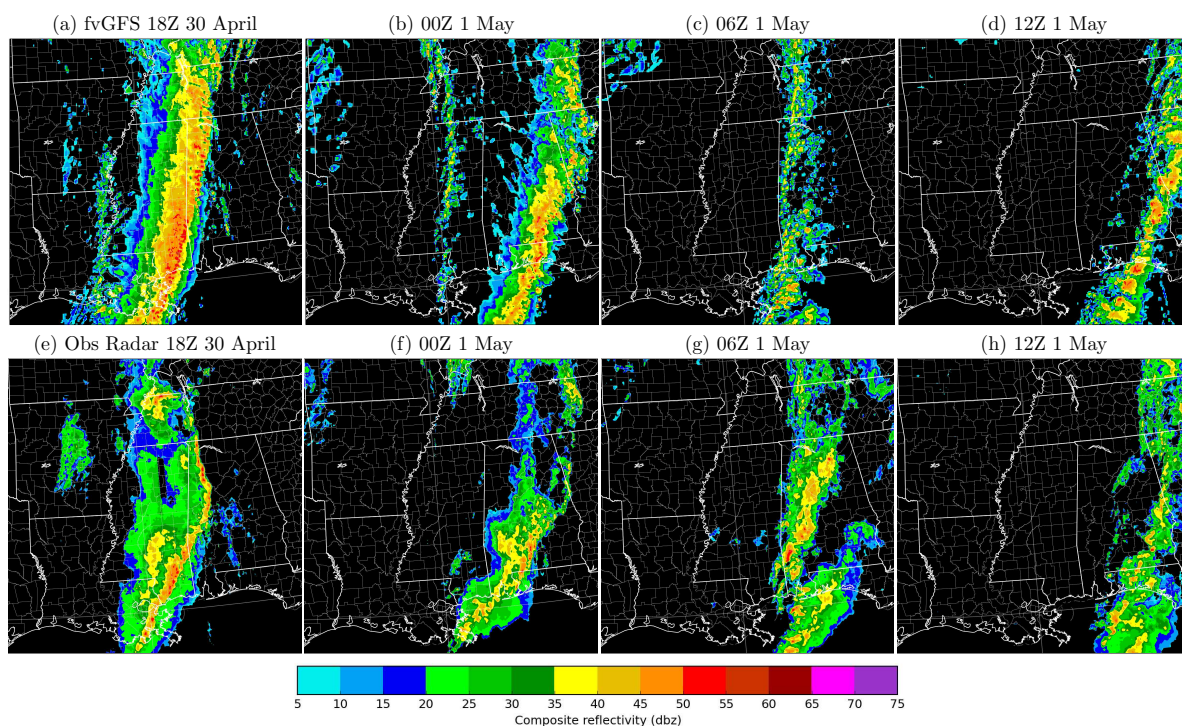
275 **Figure 11.** Time-longitude diagram of predicted hourly-accumulated precipitation (mm) averaged from 30
 276 to 50 latitude and of forecasts for the period of 1–31 May 2017, showing lead times from 12 to 48 hours: (a)
 277 13-km fvGFS; (b) nested fvGFS. (c) As in (a) and (b) but for observed Stage IV precipitation over the same
 278 time period, and showing one and a half cycles of the data. (d) Average precipitation (mm/hr) from 30 to 50
 279 latitude and 110 to 90 west longitude, for 13-km (green) and nested fvGFS (blue) and Stage IV (black). (e) As
 280 in (d) but averaged from 90 to 70 west longitude.

426 rectly predicted from Missouri into Arkansas, and propagating eastward, up to five days in ad-
 427 vance of the event. However, the squall lines propagate much more slowly the observed dere-
 428 chos or the predicted derechos in the nested fvGFS. While we expect that an environment fa-
 429 vorable for derechos is predictable several days in advance, the 13-km fvGFS cannot be ex-
 430 pected to represent the rapid propagation characteristic of derechos, owing to the lower res-
 431 olution and the parameterized nature of the convection. This case demonstrates the potential
 432 of a global-to-regional model for multi-day *explicit* prediction of severe convective systems
 433 not possible with existing global prediction models.

440 5 Conclusion and prospects for a unified global-to-regional model

441 Regionally-refined global models show great promise for extending the convective-scale
 442 prediction capability of regional models from a single day to multiple days and to medium-
 443 range forecast scales, but present many scientific challenges not arising on regional scales. The
 444 nested global-to-regional version of fvGFS we have described for convective-scale prediction,
 445 through its stability, efficiency, and forecast skill, is capable of both excellent global-scale pre-
 446 diction, meeting or outperforming the operational GFS, while also showing the ability to ex-
 447 plicitly predict individual convective storms up to five days in advance, and with precipitation
 448 skill comparable to operational convective-scale models. This explicit multi-day prediction would
 449 not be possible without skillful prediction of the global scales, which force and provide the
 450 predictability for the storm events, and the capacity for the model to represent marginally-resolved
 451 convective clouds. The fvGFS model is able to combine a skillful global prediction system
 452 and realistic simulation of storm-scale features into a single model, in the same forecast.

453 The introduction of a two-way nest into the global model does not disrupt the predic-
 454 tion skill of the large-scale circulation. The nested fvGFS shows only a minor reduction in the
 455 hemispheric 500-mb skill compared to the 13-km fvGFS, and after a short equilibration pe-
 456 riod still has a higher skill than the operational GFS, especially at longer lead times (5–7 days).
 457 The 3-km nest shows robust improvement in temperature and humidity biases compared to lower-



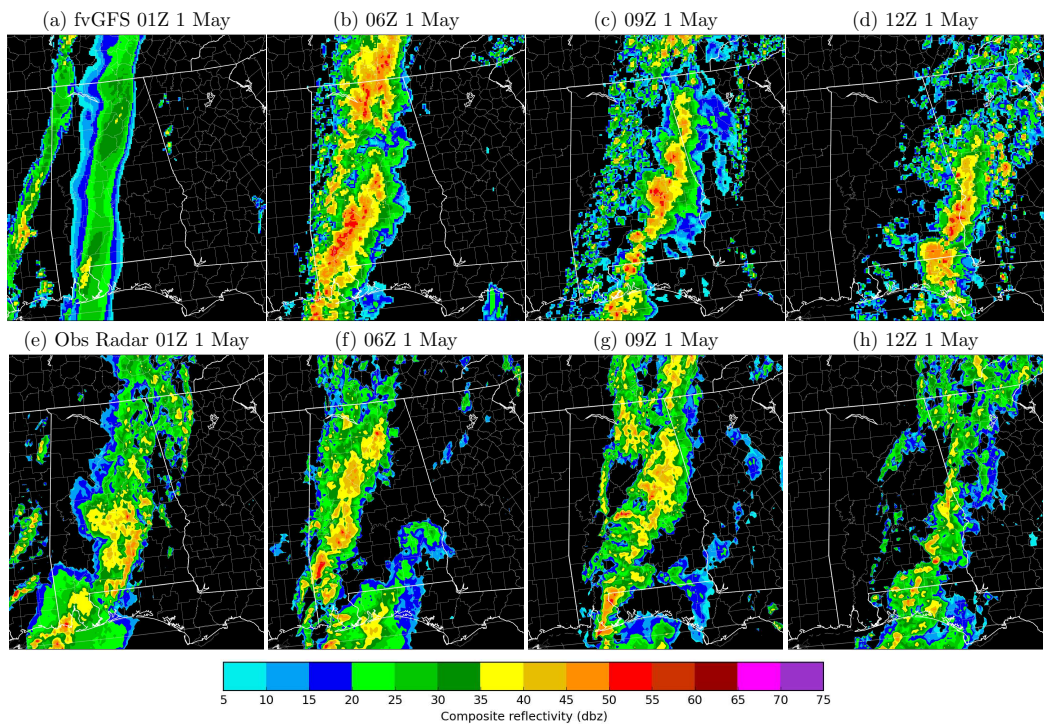
300 **Figure 12.** Composite reflectivity from fvGFS forecast initialized 00Z 30 April 2017 (top row, a–d) and
 301 observations (bottom row, e–h). Henceforth, all 3-km model output depicts shaded unsmoothed native nested-
 302 grid cells, unless otherwise stated.

458 resolution models, especially over the complex terrain of the western US. The 3-km nest also
 459 greatly improves precipitation biases, especially the low bias in extreme precipitation rates seen
 460 in the 13-km models, and is able to reproduce the propagation of individual convective sys-
 461 tems. Precipitation skill in the nested fvGFS is competitive with the operational 3-km NAM
 462 nest, as are temperature and humidity biases in the Western US, although overall temperature
 463 and humidity errors are slightly higher.

464 Individual case evaluations focused on the depiction of warm-season severe weather events.
 465 Effective prediction of frontal squall lines, isolated single-cell and supercell storms, and of dere-
 466 choes was demonstrated. These examples were used to illustrate some of the characteristics
 467 of fvGFS, including the simulated diurnal cycle and multi-day propagation of synoptic- and
 468 mesoscale precipitation systems, the spin-up from coarse-resolution initial conditions, the fine-
 469 scale structure of simulated storms (especially with respect to the representation of vorticity
 470 and updraft helicity) the source of multi-day predictability of favorable environments for se-
 471 vere convection, and the enhanced prediction value provided by refinement to convective scales.

472 The global-to-regional fvGFS configuration described in this paper has been tested and
 473 run on several NOAA supercomputing facilities. On Gaea-c4, using 1536 cores for the global
 474 grid and 1890 cores for the nested grid, the peak performance is about 18 min d^{-1} , or about
 475 twice the operational requirement of 8.5 min d^{-1} . A configuration of this unified global- and
 476 convective-scale model able to meet operational forecast needs is thereby well within the reach
 477 of current National Weather Service computing.

478 These results clearly show the promise of fvGFS for global-to-regional interactive mod-
 479 eling for improving predictions of severe weather events multiple days out, and the potential
 480 for prediction on seasonal-to-subseasonal prediction and for decadal-to-centennial climate pro-



315 **Figure 13.** As in Figure 12, except (a–d) row is fvGFS forecasts initialized 01 May 2017. Note that (a)
 316 represents one hour after initialization, and so on.

481 jection. Further work at GFDL and with our university and agency partners is underway to
 482 improve fvGFS and similar models for prediction at all temporal and spatial scales, and to im-
 483 prove the physics, initialization, and the overall configuration for a wide range of forecast and
 484 research applications.

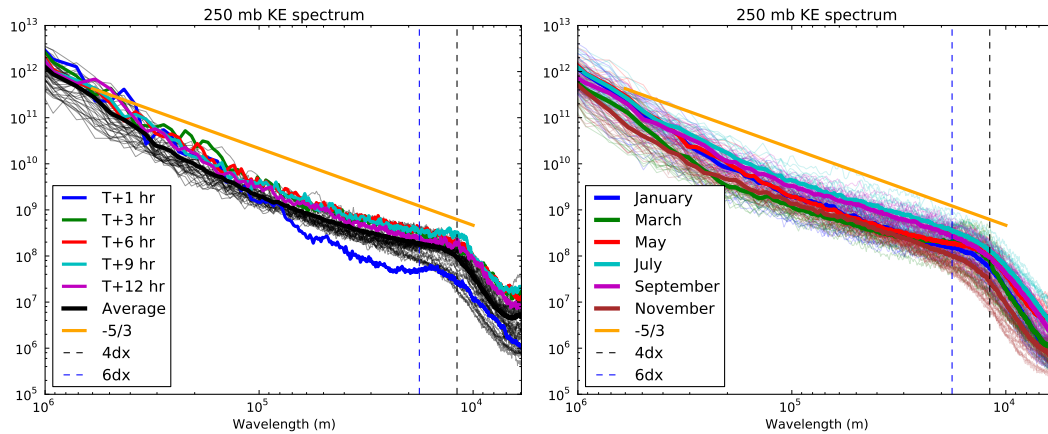
485 The results presented here make several strong points, potentially contrary to prevail-
 486 ing opinion. We have shown that an FV³-based model can produce skillful forecasts and re-
 487 alistic simulation of convective-scale features. We have also shown that the GFS physics, when
 488 appropriately modified (especially with the introduction of a modern microphysical parame-
 489 terization), is a powerful foundation for both convective-scale and global prediction. Finally,
 490 *we have shown that it is possible for a variable-resolution model to simultaneously produce skill-*
 491 *ful forecasts at every scale, from the hemispheric circulation down to convective scales.*

492 **A: Nonhydrostatic nesting in FV³**

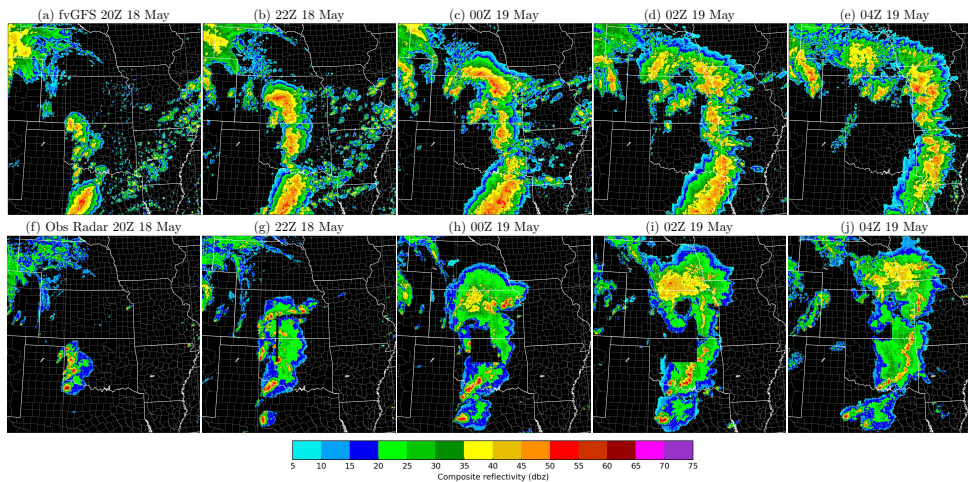
493 We describe here only the modifications to the procedure in Harris and Lin (2013).

494 The nested-grid boundary conditions for all prognostic variables, including nonhydro-
 495 static vertical velocity and density, are linearly interpolated into the outermost halo (ghost) cells
 496 of the nested grid. Boundary conditions are updated every acoustic timestep by linearly in-
 497 terpolating from two earlier coarse-grid states; for positive-definite variables the extrapolation
 498 is limited so that the minimum value is non-negative. To compute the nonhydrostatic compo-
 499 nent of the pressure on the nested-grid boundary, the pressure is diagnosed using the same semi-
 500 implicit solver used in the interior for handling the fast-time scale vertically-propagating waves.

501 Two-way updating (nested-to-coarse communication) is always performed after the physics
 502 is called. The only fields that are updated to the coarse grid are the temperature, vertical ve-



343 **Figure 14.** Nested-grid 250 mb kinetic energy spectra ($\text{m}^3 \text{s}^{-2}$). Left: forecast initialized 01 May 2017.
 344 Light gray lines are plotted every three hours starting at 15Z on 01 May; the average of these times is shown
 345 as a heavy black line. Right: Time-averaged 250-mb kinetic energy spectra for six different forecasts at
 346 different times of the year.

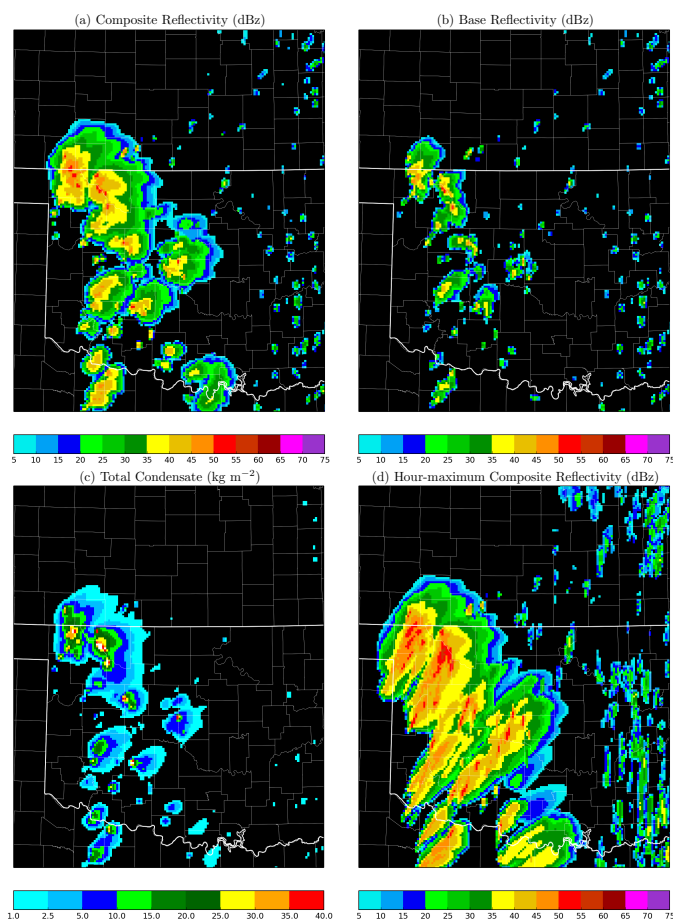


360 **Figure 15.** As in Figure 12 but for initialization at 00Z 18 May 2017.

503 locity, and staggered horizontal winds. Two-way updating of temperature and vertical veloc-
 504 ity is done using the averaging procedure in Harris and Lin (2013), whereas the update of the
 505 horizontal winds uses the vorticity-conserving procedure described in the same paper.

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 509 We also thank Rusty Benson, Zhi Liang, Jan-Huey Chen, Xi Chen, and Kun Gao for their con-
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 514 the manuscript. We also thank Ming Xue, Eric Aligo, and the participants in the Hazardous
 515 Weather Testbed's Spring Forecasting Experiment for their feedback and advice. The partic-

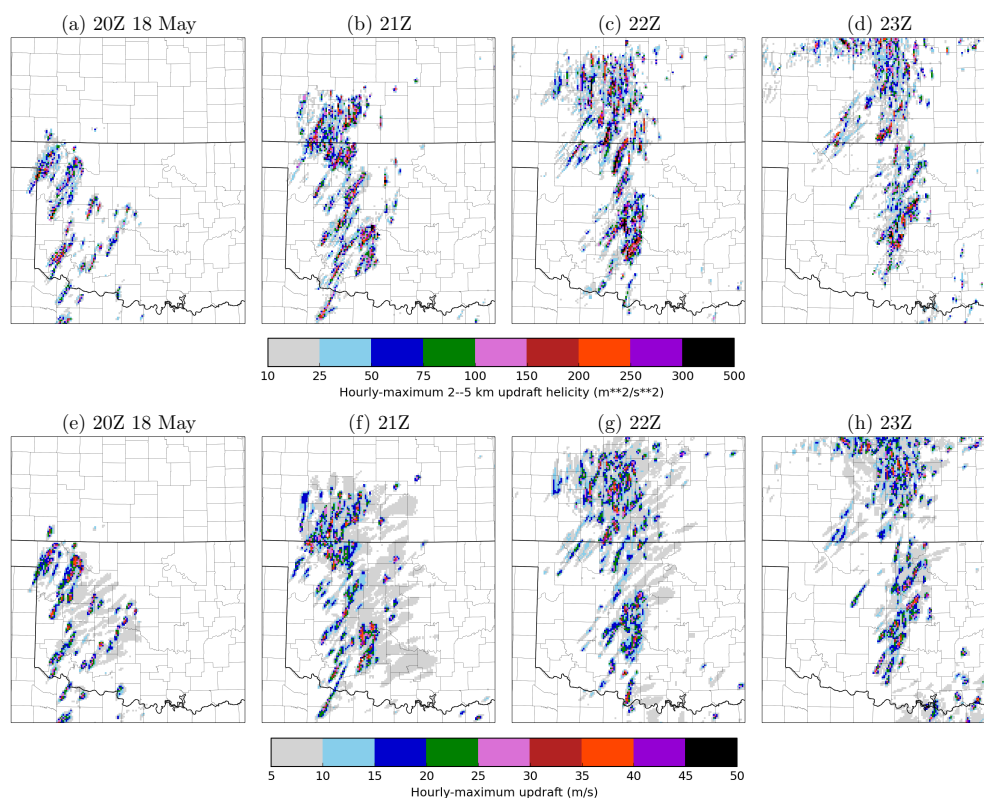


370 **Figure 16.** Nested fvGFS fields of a 20-hour forecast valid 20Z 18 May 2017. (a) Composite (column-
 371 maximum) reflectivity; (b) base (1-km above ground level) reflectivity; (c) Column integrated condensate; (d)
 372 Hourly-maximum composite reflectivity over the period 19Z to 20Z.

516 ipation of Morin, Rees, Zhou, and Stern was funded through the Next Generation Global Pre-
 517 diction System project from the NWS Office of Science and Technology Integration. All codes
 518 used for the simulations described in this paper are preserved in GFDL's GitLab version-control
 519 repository, and all data files for the simulations described in this paper are archived in the GFDL
 520 Tape Archive System; versions and data storage locations are described in the Supplemental
 521 Information.

522 References

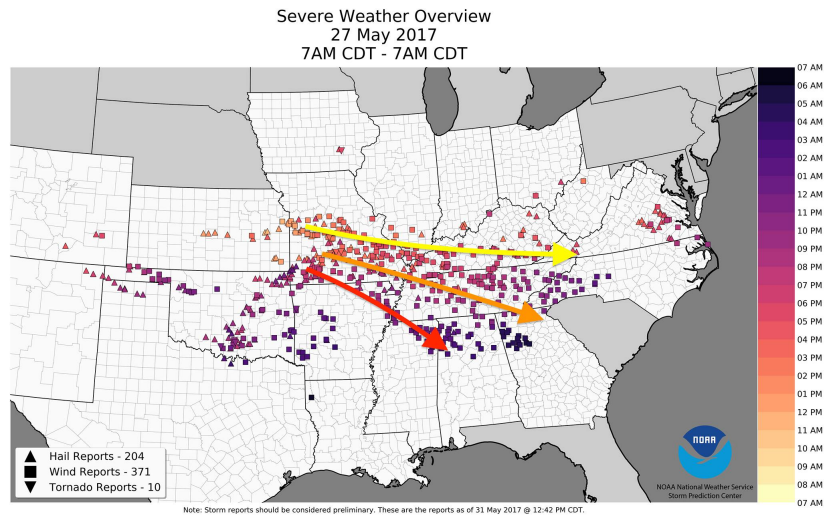
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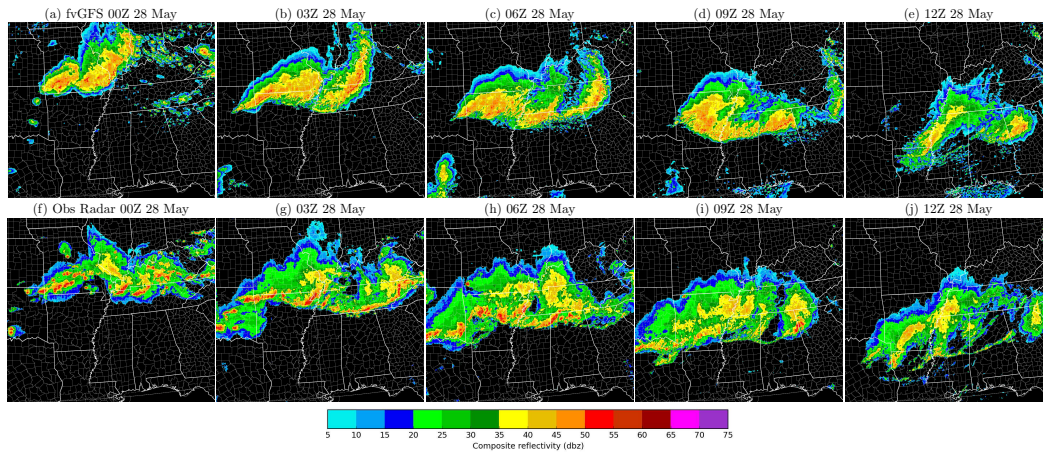
390 **Figure 17.** Nested fvGFS plots of hourly-maximum quantities of a forecast initialized 00Z 18 May 2017.
 391 (a–d): 2–5 km updraft helicity; (e–f): column-maximum updraft.

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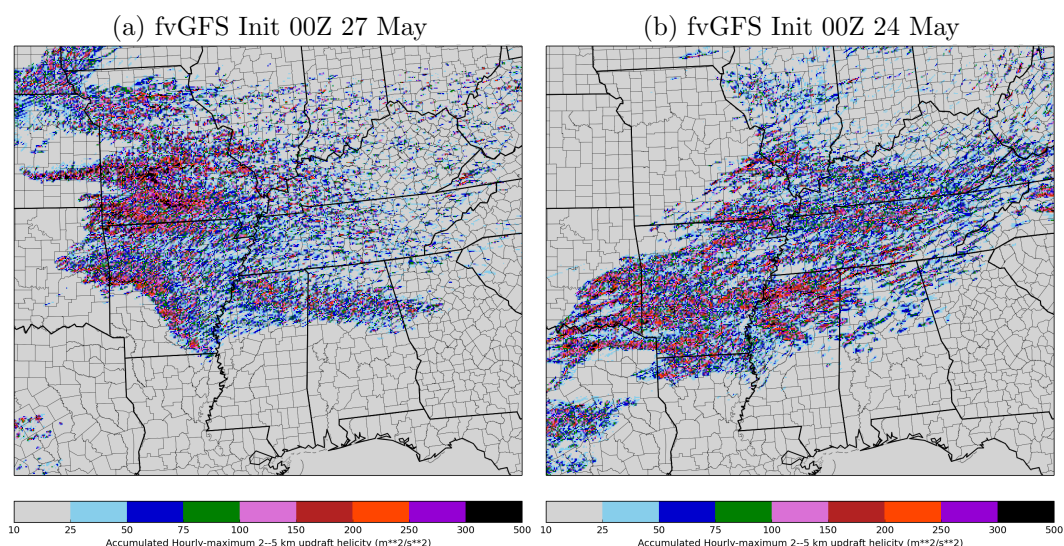


405 **Figure 18.** SPC preliminary storm reports for the 24 hours ending 12Z 28 May 2017, with subjectively-
406 analyzed derecho tracks. (Plot courtesy SPC.)



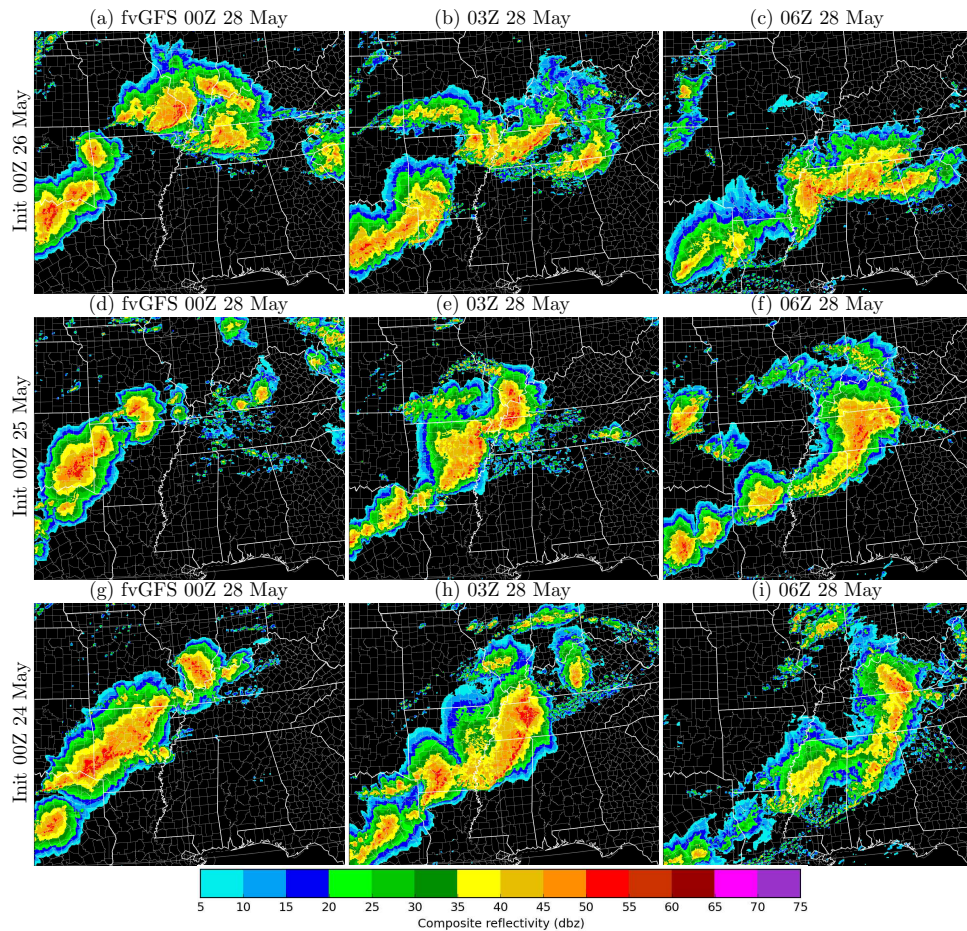
407 **Figure 19.** As in Figure 12, but for the forecast initialized 00Z 27 May 17.

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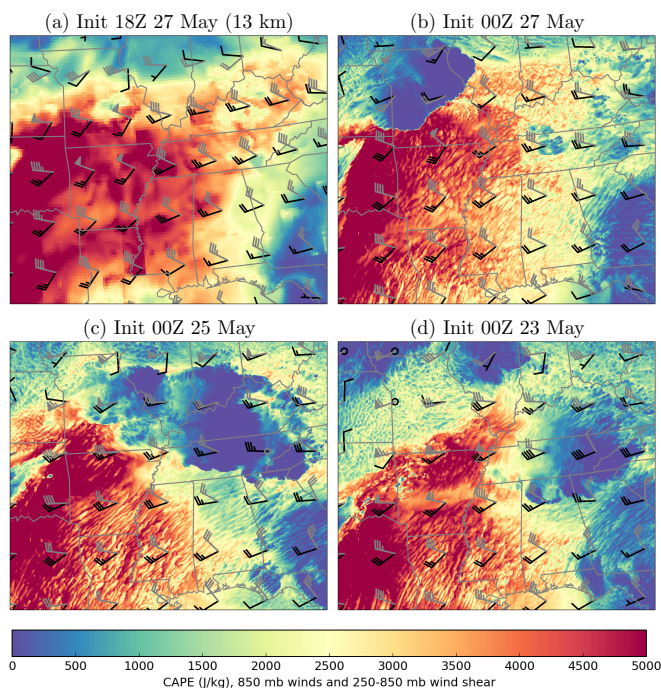
408 **Figure 20.** Maximum 2–5 km updraft helicity between 12Z on 27 May and 12Z on 28 May for fvGFS
 409 forecasts initialized (a) 00Z 27 May 17 and (b) 00Z 24 May 17.

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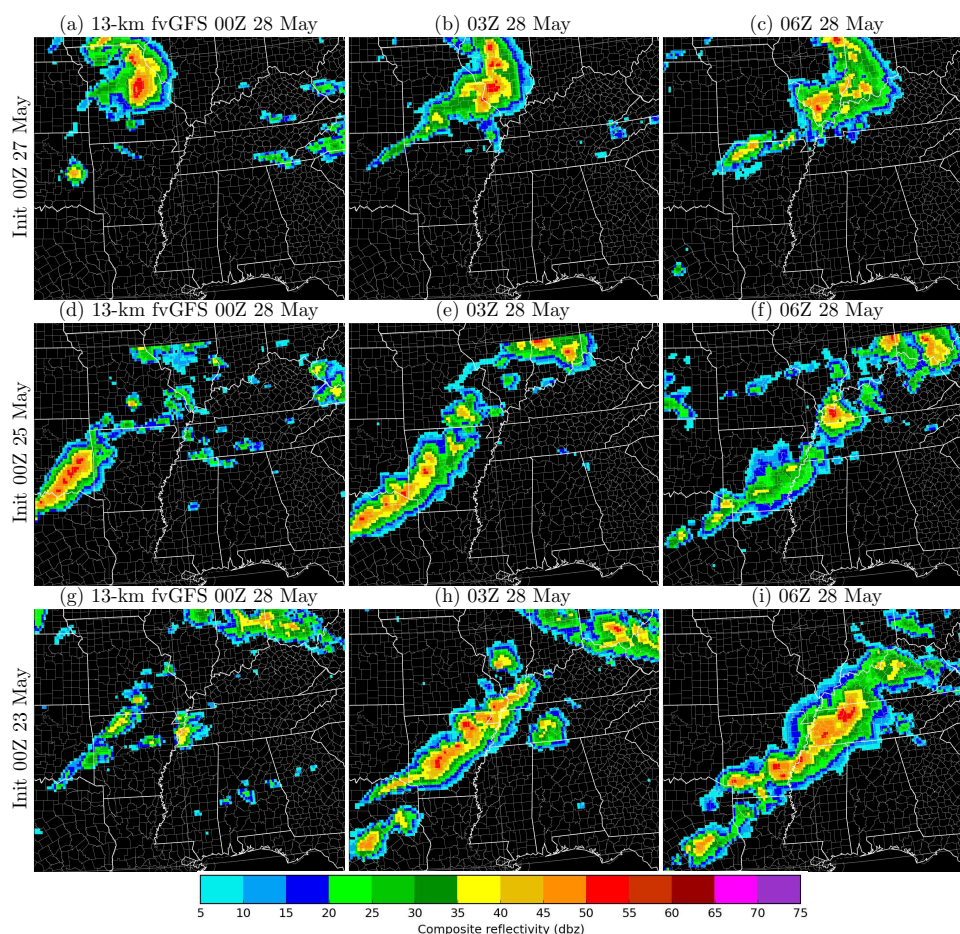
418 **Figure 21.** Multi-day forecasts of the 27 May event. (a–c): initialization 00Z 26 May. (d–f): initialization
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434 **Figure 22.** Convective available potential energy (shading), 850-mb winds (black wind barbs) and 250-mb
 435 to 850-mb wind shear (gray) valid 21Z 27 May 2017: (a) 13-km fvGFS initialized 18Z on 27 May 2017, a
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438 **Figure 23.** As in Figure 21, but for the 13-km fvGFS and for initializations of (a–c) 00Z 27 May, (d–f) 00Z
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