

The Impact of Large-Scale Orography on Northern Hemisphere Winter Synoptic Temperature Variability

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ABSTRACT

14 The impact of large-scale orography on wintertime near-surface (850hPa)
15 temperature variability on daily and synoptic time-scales (days to weeks) in
16 the Northern Hemisphere is investigated. Using a combination of theory,
17 idealized modeling work and simulations with a comprehensive climate
18 model, it is shown that large-scale orography reduces upstream temperature
19 gradients, in turn reducing upstream temperature variability, and enhances
20 downstream temperature gradients, enhancing downstream temperature vari-
21 ability. Hence the presence of the Rockies on the western edge of the North
22 American continent increases temperature gradients over North America
23 and, consequently, increases North American temperature variability. By
24 contrast, the presence of the Tibetan Plateau and the Himalayas on the
25 eastern edge of the Eurasian continent damps temperature variability over
26 most of Eurasia. However, Tibet and the Himalayas also interfere with the
27 downstream development of storms in the North Pacific storm track, and thus
28 damp temperature variability over North America, by approximately as much
29 as the Rockies enhance it.

30

31 Large-scale orography is also shown to impact the skewness of down-
32 stream temperature distributions, as temperatures to the north of the enhanced
33 temperature gradients are more positively skewed while temperatures to the
34 south are more negatively skewed. This effect is most clearly seen in the
35 northwest Pacific, off the east coast of Japan.

³⁶ **1. Introduction**

³⁷ Temperature variability is one of the most important features of the climate for human society
³⁸ and natural ecosystems, affecting, among many other things, agricultural and economic produc-
³⁹ tion (Lazo et al. (2011); Wheeler and von Braun (2013); Shi et al. (2015); Jahn (2015); Bathi-
⁴⁰ any et al. (2018)) and the rhythms of ecological seasons (Jackson et al. (2009); Bowers et al.
⁴¹ (2016)). Changes in temperature variability may be among the most impactful aspects of future
⁴² climate change, which has motivated much recent work on the mechanisms controlling temper-
⁴³ ature variability in present and future climates, with two primary foci: (1) the question of how
⁴⁴ Arctic amplification will influence mid-latitude temperature variability; and (2) the question of
⁴⁵ what controls the zonal-mean variance and higher-order moments of the temperature distribution
⁴⁶ (e.g., Schneider et al. (2015); Garfinkel and Harnik (2017); Linz et al. (2018)). With respect to
⁴⁷ Arctic amplification, it is now clear that, in winter, mid-latitude zonal-mean temperature variance
⁴⁸ will be reduced (Screen (2014); Schneider et al. (2015); Hoskins and Woollings (2015)), though
⁴⁹ the effect of Arctic amplification on higher moments of mid-latitude temperature distributions is
⁵⁰ still uncertain (e.g., Cohen et al. (2014); Barnes and Polvani (2015)).

⁵¹ Little work, however, has been done to understand what controls regional (zonally-asymmetric)
⁵² patterns of temperature variability, despite their societal relevance. For instance, changes in heat-
⁵³ waves with global warming can be well predicted by superposing a mean shift on present-day
⁵⁴ daily temperature variability, so that understanding the pattern of temperature variance is key for
⁵⁵ forecasting spatial variations in heat-wave changes with warming (Rahmstorf and Coumou (2011);
⁵⁶ Lau and Nath (2012); Lau and Nath (2014); Huybers et al. (2014); McKinnon et al. (2016)).

⁵⁷ An example of a regional difference in temperature variability can be seen in panels a and b
⁵⁸ of Figure 1. Whether using daily data (panel a) or filtering to synoptic time-scales (days to

59 weeks, panel b), North America experiences substantially more near-surface (850hPa) tempera-
60 ture variability than Eurasia during boreal winter (December-January-February, DJF, see section
61 2 for description of observational dataset). This is also shown by Figure 1e, which plots a longi-
62 tudinal profile of DJF synoptic temperature variance at 50°N: temperature variance at this latitude
63 is roughly twice as large over North America as over Eurasia. Investigating the contribution of
64 large-scale Northern Hemisphere orography (Asian orography, which includes the Himalayas, the
65 Tibetan Plateau and the Mongolian Plateau, and the Rockies) to the enhancement of temperature
66 variability over North America compared to Eurasia is the primary goal of the present study.

67 Our analysis is based on the dominant control of winter synoptic temperature variability by hor-
68 izontal advection, which implies in turn that mean horizontal temperature gradients, particularly
69 meridional gradients, are the primary control on synoptic temperature variability (Schneider et al.
70 (2015); Holmes et al. (2016); see section 3a below). It can be seen in panels c and d of Figure 1 that
71 both zonal and meridional temperature gradients are larger over North America than over Eurasia
72 during winter, suggesting that whatever causes these enhanced gradients is also responsible for the
73 enhanced variability over North America. Specifically, the importance of temperature gradients
74 for synoptic temperature variability implies a close link between the Northern Hemisphere winter
75 stationary wave pattern and the regional distribution of winter temperature variability.

76 Waves forced by large-scale orography are a key component of the winter stationary wave pattern
77 in the Northern Hemisphere (Held et al. 2002). Below, we show that orography increases down-
78 stream temperature gradients and decreases upstream temperature gradients, with corresponding
79 impacts on temperature variability. We demonstrate this mechanism in simulations with two ideal-
80 ized atmospheric general circulation models (GCMs), one dry and one moist, which also allow us
81 to investigate how the shape of the orography influences its impact on temperature variability and
82 how moist processes impact the dynamics (section 3). We then present simulations with a com-

83 prehensive climate model in which the major Northern Hemisphere mountain ranges are flattened,
84 to quantify the impact these have on winter temperature variability (section 4). A complicating
85 factor is orography’s effect on downstream development: the presence of large-scale orography
86 can weaken downstream eddies by interfering with the recycling of energy from upstream, leading
87 to reduced temperature variability far from the orography.

88 By enhancing and reducing mean temperature gradients, orography also impacts the skewness
89 of temperature distributions, which we explore in section 5. We end with conclusions in section 6.

90 **2. Data and Methods**

91 *a. Observational Data*

92 Observational data are taken from the Modern-Era Retrospective Analysis for Research and
93 Applications (MERRA) dataset (Rienecker et al. 2011). The MERRA grid has 1.25° resolution
94 in latitude and longitude, and we have taken daily-averaged data from December, January and
95 February for the years 1979 to 2012.

96 *b. Dry GCM*

97 The dry GCM is the GFDL spectral dynamical core, which solves the primitive equations for
98 a dry ideal gas on the sphere, and is forced by Newtonian relaxation to a prescribed zonally-
99 symmetric equilibrium temperature field and damped by Rayleigh friction near the surface. The
100 parameter settings are the standard Held-Suarez parameters with forcing symmetric about the
101 equator (Held and Suarez 1994). This set-up produces an equinoctial climate similar to that of the
102 real atmosphere, though there are no stratospheric polar vortices due to the uniform stratospheric
103 relaxation temperature.

104 As in Lutsko and Held (2016), the model is perturbed by adding a Gaussian mountain, with the
105 form

$$h(\phi, \lambda) = H \exp \left\{ - \left[\frac{(\phi - \phi_0)^2}{\alpha^2} + \frac{(\lambda - \lambda_0)^2}{\beta^2} \right] \right\}, \quad (1)$$

106 where H is the maximum height of the mountain in meters; λ and ϕ are longitude and latitude,
107 respectively; λ_0 and ϕ_0 are the co-ordinates of the center of the mountain; and α and β are half-
108 widths, both set to 15° in the main suite of simulations. λ_0 and ϕ_0 were set to 90°E and 45°N ,
109 respectively, in all simulations.

110 H was varied from 333m, which is in the “linear” regime, with air mostly flowing up and over
111 the mountain, to 4km, which is in the “non-linear” regime, with air mostly deflected around that
112 orography (Lutsko and Held 2016). In every simulation, the model was run at T85 resolution with
113 30 evenly spaced sigma levels, and the instantaneous wind, surface pressure and temperature fields
114 were sampled once per day. We present results from simulations lasting 5000 days, with data taken
115 from the final 4000 days.

116 *c. Moist GCM*

117 The moist GCM is the gray-radiation model first described by Frierson et al. (2006), though
118 we have used the parameter settings of O’Gorman and Schneider (2008), and also included their
119 parameterization of short-wave absorption by the atmosphere. The model uses the GFDL spectral
120 dynamical core, and includes the simplified Betts-Miller (SBM) convection scheme of Frierson
121 (2007). We show results using a convective relaxation time-scale τ_{SBM} of 2 hours and a refer-
122 ence relative humidity $RH_{\text{SBM}} = 0.7$. The boundary layer scheme is the one used by O’Gorman
123 and Schneider (2008). The moist GCM is run under perpetual equinox conditions, with no daily
124 cycle of insolation, and is coupled to a slab ocean of depth 1m, with no representation of ocean

125 dynamics or of sea ice. A mixed-layer depth of 1m was used so that the model would spin up
126 quickly; using a deeper mixed-layer damps the temperature variance, but otherwise our results
127 are qualitatively insensitive to the choice of mixed-layer depth. Moreover, a mixed-layer depth of
128 1m allows surface temperatures to respond to synoptic-scale forcing, as continental land surfaces
129 do. A deeper mixed-layer depth, more representative of an oceanic mixed-layer, would decouple
130 surface temperatures from synoptic temperature variability.

131 The same Gaussian orography is added to the model as in the dry GCM, except that it is centered
132 further north at 60°N. The reason for moving the orography poleward is that the storm tracks, and
133 the associated maxima in temperature variance, are further poleward in this set-up (see Figure 3
134 below), so a more northward mountain produces clearer changes in variance. As discussed by
135 Wills and Schneider (2018), this implementation of orography produces an “aqua-mountain”, and
136 the surface fluxes over the orography are not necessarily realistic. However, any bias in the surface
137 fluxes is of secondary importance for our investigation.

138 The moist GCM was integrated at T85 truncation with 30 unevenly-spaced vertical levels, start-
139 ing from a state with uniform SSTs. The simulations lasted for 4500 days with data stored four
140 times per day, and we have taken averages over the final 4000 days.

141 Our focus in this study is on winter temperature variability, as land surface processes, like soil-
142 moisture feedbacks, are less important for variability in winter than in summer. As neither of the
143 idealized GCMs includes a representation of land surface processes, they can be used to study the
144 mechanisms of winter temperature variance without imposing seasonality and so, for convenience,
145 we have used set-ups that produce equinoctial climates.

¹⁴⁶ *d. Comprehensive climate model*

¹⁴⁷ The comprehensive climate model is GFDL CM2.5-FLOR (Vecchi et al. 2014). FLOR stands for
¹⁴⁸ Forecast-oriented Low Ocean Resolution, and the model is based on the GFDL CM2.5 model. It
¹⁴⁹ is run with an atmospheric resolution of approximately 50km and an oceanic resolution of approx-
¹⁵⁰ imately 1°. By running with a relatively high resolution atmosphere, FLOR is able to accurately
¹⁵¹ capture many subseasonal forms of variability, such as hurricanes and monsoon depressions, and
¹⁵² can resolve sharp topographic features, such as the peaks of the Himalayas (compare panels a and
¹⁵³ b of Figure 2).

¹⁵⁴ Three simulations were performed with FLOR: (1) a control simulation with present-day to-
¹⁵⁵ pography, (2) a simulation with the Rockies flattened to 300m (the “no-Rockies” simulation, i.e.,
¹⁵⁶ all surface heights greater than 300m are reduced to 300m) and (3) a simulation with the Asian
¹⁵⁷ orography (the Tibetan Plateau, the Himalayas and the Mongolian Plateau) flattened to 300m (the
¹⁵⁸ “no-Tibet” simulation). The regions of flattened topography can be seen in Figure 2 and we note
¹⁵⁹ that the gravity wave drag and boundary layer roughness were fixed to their control values where
¹⁶⁰ the topography was flattened (see also Baldwin et al. (2019b)).

¹⁶¹ All simulations were conducted with pre-industrial radiative forcings, matching the best guess
¹⁶² for the year 1860, and with static vegetation. Daily-mean data were collected for 50 years, fol-
¹⁶³ lowing 100 years of spin-up from an initial state of rest, and SSTs were relaxed to a repeating
¹⁶⁴ climatology with a relaxation time-scale of five days. This set-up was originally designed to al-
¹⁶⁵ low tropical cyclones to interact with the ocean surface (Vecchi et al. 2014); for our purposes, the
¹⁶⁶ model is essentially an atmosphere-only climate model run over fixed SSTs. Our configuration
¹⁶⁷ attempts to isolate the direct effects of the orographic forcing on temperature variability, though
¹⁶⁸ not the indirect effects orography has on variability through its impact on SSTs.

¹⁶⁹ *e. Filtering to synoptic time-scales*

¹⁷⁰ The data were filtered to synoptic time-scales using a fourth-order Butterworth filter, with cut-
¹⁷¹ off frequencies of $1/3$ days $^{-1}$ and $1/15$ days $^{-1}$. The filter was implemented using the Python
¹⁷² package `scipy.signal`, with the filter co-efficients obtained using `scipy.signal.butter` and
¹⁷³ the filtering done with `scipy.signal.lfilter`. We have verified that our results are robust to
¹⁷⁴ the choice of filtering time-scales, within reason. For all datasets, DJF variance and skewness were
¹⁷⁵ calculated individually for each year (e.g., December 1979 to February 1980) and then averaged
¹⁷⁶ over all years to find the climatological variance and skewness.

¹⁷⁷ **3. Impact of Orography on Temperature Variance in Idealized Models**

¹⁷⁸ *a. Background theory*

¹⁷⁹ Assuming that synoptic potential temperature variations are primarily generated by horizontal
¹⁸⁰ advection, and that this advection is local in time and space, potential temperature variations can
¹⁸¹ be Taylor expanded to give (Corrsin (1974); Schneider et al. (2015))

$$\theta' \approx -\frac{\partial \bar{\theta}}{\partial y} L'_y - \frac{\partial \bar{\theta}}{\partial x} L'_x + \frac{1}{2} \frac{\partial^2 \bar{\theta}}{\partial y^2} L'^2_y + \dots, \quad (2)$$

¹⁸² where $\theta' = \theta - \bar{\theta}$ denotes synoptic variations of potential temperature at 850hPa about some lo-
¹⁸³ cal mean value $\bar{\theta}$, L'_y is the Lagrangian displacement of air masses arriving at y from y_0 , and
¹⁸⁴ similarly for L'_x . “Mean” denotes an average over a time-scale that is long compared to synoptic
¹⁸⁵ time-scales and we consider potential temperature rather than temperature because potential tem-
¹⁸⁶ perature is materially conserved during adiabatic airmass displacements. We work in Cartesian
¹⁸⁷ co-ordinates for simplicity, and define L'_y as positive for a northward displacement and L'_x as posi-
¹⁸⁸ tive for an eastward displacement. Provided the length scales of potential temperature variations,
¹⁸⁹ $\overline{L_y} = 2|\partial_y \bar{\theta} / \partial_{yy} \bar{\theta}|$ and $\overline{L_x} = 2|\partial_x \bar{\theta} / \partial_{xx} \bar{\theta}|$, are much larger than the mixing length-scales, L'_y and L'_x ,

190 the expansion can be well approximated by just retaining the first two terms, and so the synoptic
 191 potential temperature variance can be approximated as

$$\overline{\theta'^2} \approx \overline{\left(\frac{\partial \bar{\theta}}{\partial y}\right)^2 L_y'^2} + \overline{\left(\frac{\partial \bar{\theta}}{\partial x}\right)^2 L_x'^2} + 2 \overline{\left(\frac{\partial \bar{\theta}}{\partial y}\right) L_y' \left(\frac{\partial \bar{\theta}}{\partial x}\right) L_x'}. \quad (3)$$

192 The meridional term, specifically the meridional temperature gradient, generally dominates over
 193 the zonal term and the cross term (note the different colorbar scales in panels c and d of Figure 1),
 194 but we have included the latter two here to emphasize that zonal temperature gradients also impact
 195 regional potential temperature variability.

196 Orography affects temperature gradients by meridionally compressing downstream near-surface
 197 isentropes and pulling apart upstream isentropes. However, this requires the flow to be deflected
 198 around the orography, rather than flowing up and over it, so that the air deflected equatorward
 199 partly adjusts to the warmer conditions and the air deflected poleward partly adjusts to the colder
 200 conditions, before the downstream confluence of the flow. For small heights the air flows up and
 201 over the orography, leaving the potential temperature gradients unaffected. Formally, consider the
 202 linearized, time-mean thermodynamic equation for adiabatic flow on the lowest model level (in z
 203 coordinates):

$$\bar{u} \frac{\partial \theta'}{\partial x} + v' \frac{\partial \bar{\theta}}{\partial y} = -w \frac{\partial \bar{\theta}}{\partial z}. \quad (4)$$

204 The orographic forcing enters through the lower boundary condition, which can be approximated
 205 in the linear regime as (Cook and Held 1992)

$$w \approx \bar{u} \frac{\partial h}{\partial x}, \quad (5)$$

206 where h is again the height of the orography, with maximum height H . Substituting then gives

$$\bar{u} \frac{\partial \theta'}{\partial x} + v' \frac{\partial \bar{\theta}}{\partial y} = -\bar{u} \frac{\partial h}{\partial x} \frac{\partial \bar{\theta}}{\partial z}. \quad (6)$$

207 In this linear regime, the air moves up and over the mountain, and the first term on the left side of
 208 equation 6 balances the right side so that $\frac{\theta'}{H} \sim \frac{\partial \bar{\theta}}{\partial z}$. For larger H the flow becomes increasingly non-
 209 linear, and the deflection around the orography is important. In this regime the forcings associated
 210 with the meridional wind and with zonal wind anomalies can no longer be ignored in equation 5,
 211 and the orographic forcing is balanced by the $v' \frac{\partial \bar{\theta}}{\partial y}$ term¹. In idealized experiments this transition
 212 occurs for H between 1 and 2km for orography with approximately the same horizontal extent as
 213 the Tibetan Plateau (Cook and Held (1992), Lutsko and Held (2016)). Valdes and Hoskins (1991)
 214 demonstrated that Asian topography meets this criterion, but caution that it is less clear whether
 215 the Rockies do, with the result depending on how the Rockies are defined. Furthermore, while
 216 the near-surface flow appears to be deflected around the Rockies (see Figure 9 below), this flow
 217 is strongly influenced by heating in the north Pacific storm track (see also Valdes and Hoskins
 218 (1989)).

219 Another factor which enhances the downstream temperature gradients is the preferential deflec-
 220 tion of the flow around the poleward side of the orography. If the flow follows isentropes, then
 221 it will descend in height when it moves equatorward and ascend in height when it moves pole-
 222 ward, following the mean isentropic slope (Valdes and Hoskins 1991). Thus the mountain appears
 223 “taller” to the flow on its equatorward flank and “shorter” on its poleward flank, so that more of
 224 the air flows around the poleward flank of the mountain. The downstream convergence is then
 225 equatorward of the center of the orography, with anomalously cold air meeting the warm air that
 226 flowed around the equatorward side of the orography.

¹Though note that the potential temperature perturbation is itself proportional to the deflection of the flow: $\theta' \approx \eta' \frac{\partial \bar{\theta}}{\partial y}$, where η is the typical meridional displacement of a fluid parcel, assumed to be equal to the meridional extent of the orography. Hence the condition for meridional deflection to dominate is $|\eta'/H| < \left| \frac{\partial \bar{\theta}}{\partial z} / \frac{\partial \bar{\theta}}{\partial y} \right|$. Roughly, the meridional slope of the mountain must be greater than the characteristic slope of the isentropes (Valdes and Hoskins 1991).

227 *b. Idealized GCM results*

228 In both idealized GCMs, temperature variance is reduced upstream and enhanced downstream
229 of orography (Figure 3a and b), as are meridional temperature gradients (panels c and d). However
230 the inferred mixing lengths $L' = \sqrt{\theta'^2 / \left(\frac{\partial \bar{\theta}}{\partial y} \right)^2}$ are reduced downstream of the orography (panels
231 e and f), which is the result of two competing effects. First, by increasing downstream temperature
232 gradients, orography increases downstream Eady growth rates, potentially leading to more ener-
233 getic eddies and thus to larger mixing lengths (see Caballero and Hanley (2012) for discussion of
234 the relationship between eddy kinetic energy and mixing lengths). But in addition to local baro-
235 clinicity, eddies in strong jets are also energized by downstream development – by the recycling
236 of energy from upstream eddies (Chang and Orlanski (1993); Chang et al. (2002)). Orography
237 disrupts the latter by interfering with the zonal propagation of wave packets (Son et al. 2009), and
238 for the set-ups used here this effect wins out, resulting in less energetic eddies and smaller effec-
239 tive mixing lengths. This reduction in the mixing lengths has a substantial impact on the local
240 response of the variance. For instance, panel a of Figure 4 shows the zonal anomalies in synoptic
241 temperature variance for the simulation with the dry GCM and $H = 4\text{km}$, and it can be seen that
242 the reduction in the mixing lengths creates a small region, near 120°E , in which the downstream
243 variance is reduced, while the largest increase in variance is further downstream, at around 170°E ,
244 where the eddies are more energetic.

245 On the poleward side of the mountain the pattern is reversed (Figure 4a), with enhanced temper-
246 ature variance upstream and reduced variance downstream of the mountain. This is partly caused
247 by the preferential deflection of the flow around the poleward flank of the mountain (arrows in Fig-
248 ure 3a), which induces convergence on the northwest flank of the mountain, and thus a tightening
249 of the isentropes, and divergence on the northeast flank of the mountain, causing the isentropes to

250 pull apart (see contours in Figure 3c). The jet is also relatively narrow in the dry GCM, compared
251 to typical winter climates, so that there are strong polar easterlies at the latitudes of the poleward
252 edge of the mountain. Hence the northeast flank is upstream of the mountain, and temperature
253 variance should be reduced there.

254 Our focus is on the jet regions, however, where the enhanced meridional temperature gradients
255 cause a local enhancement of temperature variance downstream of the orography in both models.
256 Figure 5 shows that the maximum zonal anomaly in potential temperature variance increases in the
257 simulations with the dry and moist GCMs as the height of the orography (H) is increased (panel
258 a)², as does the maximum zonal anomaly of the squared meridional temperature gradient (panel
259 b). Plotting these against each other demonstrates the strong linear relationship between the two
260 quantities in the GCMs (panel c). The different slopes indicate that the mixing lengths differ in the
261 two models, and the larger slope for the moist GCM implies that adding moist processes increases
262 the effective mixing length (see below).

263 A possible complication is the shape of the orography: the Rockies form a meridionally-
264 elongated ridge, whereas the Himalayas are more zonally-elongated. To investigate how the orog-
265 raphy's shape influences temperature variability, two additional simulations were run with the dry
266 GCM, one with a 4km meridional ridge resembling the Rockies ($\alpha = 15^\circ$ and $\beta = 5^\circ$) and one
267 with a 4km zonal ridge ($\alpha = 5^\circ$ and $\beta = 15^\circ$).

268 Panels b and c of Figure 4 show the zonal anomalies in temperature patterns in these simulations
269 (we note that the zonal-mean variance is lower in both of the ridge experiments than in the circular
270 experiment because the ridges interfere less with the downstream development and hence the mix-
271 ing lengths are larger than in the circular experiment). The zonal anomalies are broadly similar in

²Note that the zonal-mean variance decreases with increasing H in both models because of the increasing disruption of downstream development by the orography (not shown).

272 all three experiments, with reductions in temperature variance upstream of the mountains and en-
273 hancements downstream of the mountain and a reversed pattern at higher latitudes, however there
274 are some noticeable differences. For instance, in the zonal ridge case the reduction is mostly on
275 the southern flank of the mountain, rather than to the southwest. The meridional ridge produces a
276 similar response to the circular experiment, but a key difference is that the variance is increased on
277 the entire eastern flank of the meridional ridge. The Rockies show a similar local enhancement of
278 variance on their eastern flank (Figure 1). In the meridional ridge simulation the largest increase
279 in variance is also immediately downstream of the orography, on its southeastern flank, instead of
280 being displaced further downstream, as for the circular case. Decomposing this response into a
281 squared gradient and an inferred mixing length shows that in the meridional ridge case the tem-
282 perature gradient is more strongly increased immediately downstream of the orography, relative to
283 the reduction in the mixing length (not shown).

284 In the dry GCM, advection is the sole method of generating potential temperature variance,
285 whereas in the moist GCM covariance of anomalous latent heating and potential temperature
286 anomalies also contributes. To investigate the role of latent heat anomalies, Figure 6 shows the ad-
287 vective terms in the temperature variance budget (see equation 3 of Wilson and Williams (2006))
288 for the $H = 4\text{km}$ simulation with the moist GCM, as well as the contribution of latent heat fluctu-
289 ations to temperature variance ($\overline{\theta'Q'_L}$, Figure 6d). Latent heating enhances temperature variability
290 downstream of the orography, increasing the inferred mixing lengths diagnosed in the moist GCM
291 and partly explaining why the downstream maximum in temperature variability is closer to the
292 mountain in this GCM than in the dry GCM. This enhancement is around 20% of the advective
293 tendency, which is dominated by the $\overline{\mathbf{u}'\theta'} \cdot \nabla \bar{\theta}$ term, and comes about because the latent heating
294 fluctuations are related to the advection. For instance, anomalously warm air, originating close to

295 the surface in the tropics, will condense water as it moves poleward and rises, further enhancing
296 the temperature anomalies.

297 In summary, the results of the GCM simulations agree with the theoretical expectations from
298 the previous section, with reductions and enhancements of temperature variance caused mostly by
299 changes in meridional temperature gradients due to the presence of orography. This is complicated,
300 however, by reductions in the effective mixing lengths due to the interference of the orography with
301 downstream development. The ridge experiments with the dry GCM also demonstrated important
302 dependencies on the aspect ratio of the orography. In the case of a meridional ridge, resembling
303 the Rockies, the variance is enhanced immediately downstream of the orography, whereas with
304 a more “circular” orography the largest enhancement is further downstream. Finally, analyzing
305 the temperature variance of the moist GCM demonstrates that the contribution of latent heating
306 anomalies to temperature variance enhances the variance due to horizontal advection, as these
307 latent heating anomalies are tied to the advection itself. So we can proceed by focusing on the
308 advection, noting that latent heating enhances the effective mixing lengths.

309 **4. Temperature Variability in Simulations with Flattened Orography**

310 Figure 7a shows that FLOR is able to reproduce the main features of MERRA’s pattern of DJF
311 synoptic temperature variance³. In Figure 7b it can be seen that the effect of the Asian orography
312 is to decrease the temperature variance over most of Eurasia, as well as over the North Pacific and
313 North America, and to increase the variance over central Siberia (see also Figure 8). Notably, tem-
314 perature variance is reduced over the heavily populated southeast Asian coast, including southern
315 China, in the control simulation compared to the no-Tibet simulation. In part, this is because at

³The temperature variance is somewhat higher in the FLOR simulations than in the reanalysis, which we attribute in part to the higher resolution of FLOR’s atmospheric model compared to the reanalysis: coarse-graining the data from the control simulation to a 1.25° grid reduces the synoptic temperature variance by about 30% on average (not shown). See also Supplementary Figure 3 of Baldwin et al. (2019a).

these latitudes the zonal-winds transition from westerly to easterly and this region is upstream of the orography (Figure 9). However, the primary cause of the reduced variance is the Asian orography's interference with downstream development, which weakens the storms over southeast Asia and, especially, in the Pacific storm track (Figure 10c). The Kuroshio Extension off the east coast of Japan is the genesis region for the Pacific storm track, and the Himalayas and Tibet weaken the eddies formed over the Kuroshio because of the reduced energy from upstream, despite the increased temperature gradient in the northwest Pacific. The reduced downstream development also impacts the strength of winter storms originating in the Pacific storm track and reaching North America.

The winter stationary wave pattern over Eurasia consists of a zonally oriented dipole, with anomalous warmth over Europe and anomalous cold over east Asia (Figure 7g). The presence of Tibet cools east Asia (compare Figure 7 panels g and h), implying that the stationary wave forced by the Asian orography constructively interferes with the stationary wave excited by the land-sea contrast on Eurasia's east coast (Kaspi and Schneider (2011); Park et al. (2013)). In the absence of the Asian orography the largest Eurasian temperature gradients are at relatively low latitudes, with the maximum gradient at about 30°N (Figure 7e), whereas the mid-latitude jet, where the mixing lengths are largest, is further north. This southward displacement of the maximum temperature gradient when the orography is flattened contributes to the smaller temperature variance over Eurasia compared to North America in the no-Tibet simulation.

The Rockies act to increase the variance over most of North America, but also decrease the variance off the west coast of North America (panel c of Figure 7, Figure 8). Both the Rockies and the Asian orography increase the temperature variance over the polar regions, because their presence cools the high latitudes, increasing the zonal-mean equator-to-pole temperature gradient (Figure 10a). We have not fully diagnosed the reasons for this, but note that the mid-latitude jets

340 weaken in the presence of the mountain ranges, resulting in weaker poleward transient eddy heat
341 fluxes (Figure 10 panels b and c).

342 Table 1 quantifies the changes in temperature variance over the two continents by comparing DJF
343 synoptic temperature variance in the three FLOR simulations over a Eurasian box (40° - 120° E and
344 30° - 75° N) and over a North American box (240° - 280° E and 30° - 75° N). The areas of the Asian
345 mountains and the Rockies are masked whenever an average is taken over these boxes. Asian
346 orography reduces the variance over the Eurasian box by 1.4K^2 and over the North American box
347 by 1.3K^2 , with both of these changes statistically significant at the 95% level based on a two-sided
348 Student's t-test. The Rockies enhance the variance over the North American box by 1.3K^2 and
349 over Eurasia by 0.2K^2 , though only the change over North America is statistically significant in
350 this case.

351 These calculations suggest that the enhancement of North American temperature variability by
352 the Rockies is roughly canceled by the damping of variability due to Asian orography. The in-
353 creases and decreases in variance are sensitive to the definitions of the boxes, however, and this
354 cancellation also assumes the effects of flattening the mountain ranges individually can be linearly
355 added together. Regardless, the majority of the orography's net effect comes from the reduction
356 of Eurasian temperature variability by the Asian mountains and, in FLOR, this explains about a
357 quarter of the difference in variance over the two continents ($1.4\text{K}^2 / 5.5\text{K}^2 \approx 25\%$).

358 Our framework for explaining differences in temperature variance is based on differences in
359 mean temperature gradients, which are in turn controlled by the Northern Hemisphere stationary
360 wave pattern. So the remaining difference in temperature variance between the two continents can
361 largely be attributed to stationary waves forced by diabatic heating, which, together with orography
362 are responsible for the bulk of the Northern Hemisphere stationary wave pattern (Held et al. 2002).

Even in the no-Rockies simulation there are substantial meridional temperature gradients over North America (Figure 7f), and the stationary wave pattern over North America is similar in all three simulations, consisting of a dipole with anomalously warm temperatures off the west coast of North America and anomalously cold temperatures centered over northeast Canada (Figure 7 panels g), h) and i)). The dipole is weaker in the no-Rockies simulation, indicating that the stationary wave forced by the Rockies constructively interferes with the dipole. In this case, the pattern over North America is a combination of the stationary wave forced by the land-sea contrast between the east coast of North America and the western Atlantic (Kaspi and Schneider 2011), which cools eastern North America, and stationary waves forced by diabatic heating in the Pacific warm pool region and by thermal forcing in the extratropical Pacific (Hoskins and Karoly (1981); Valdes and Hoskins (1991); Held et al. (2002)). The latter includes the forcing due to the warm waters of the Kuroshio as well as the eddy sensible heat flux convergence in the Pacific storm track, making it difficult to separate out the relative contributions of the different thermal forcings.

5. Temperature Skewness

Through its effects on temperature gradients, orography also impacts the skewness of synoptic temperatures. Garfinkel and Harnik (2017) showed that, in mid-latitudes, synoptic temperature extremes occur when air is advected over regions with large mean meridional temperature gradients, so that temperatures poleward of these regions tend to be positively skewed and temperatures equatorward of these regions tend to be negatively skewed. By strengthening downstream temperature gradients, orography increases the positive skewness to the north of these gradients and the negative skewness to the south. This is illustrated in Figure 11, which shows maps of skewness in simulations with the two idealized GCMs, as well as the meridional temperature gradients. In

385 both cases, downstream temperatures are skewed more positively north of the enhanced tempera-
386 ture gradients and more negatively to the south of the gradients.

387 In the reanalysis data, the strongest DJF meridional temperature gradients are found in the storm
388 track regions of the west Pacific and the west Atlantic (Figure 1c). Panel a of Figure 12 shows that
389 synoptic temperatures are positively skewed in the northwest Pacific and the northwest Atlantic,
390 and negatively skewed to the south of these regions. The same patterns are seen in the control
391 simulation with FLOR (Figure 12b, note that as with the variance, we attribute the larger values
392 of skewness in part to FLOR’s higher resolution). The temperature gradient in the west Pacific is
393 reduced in the no-Tibet simulation, and comparing panels b and c of Figure 12 confirms that the
394 skewness in the northwest Pacific is also reduced in this simulation. Averaging over the region
395 35°N-50°N and 140°E to 180°E (green box in Figure 12b) gives a reduction in skewness of 31%
396 ($= (0.234 - 0.162) / 0.234$, difference significant at the 90% level) in the northwest Pacific.

397 Flattening the Rockies does not appear to affect temperature gradients in the west Atlantic (Fig-
398 ure 7f), and the skewness in the northwest Atlantic is comparable in the control and the no-Rockies
399 simulations. Over land, DJF synoptic temperatures are negatively skewed at almost all latitudes,
400 and other factors, such as land-surface feedbacks, are likely important for generating extreme
401 events.

402 **6. Conclusion**

403 In this study we have investigated the contribution of large-scale orography to the increased win-
404 tertime near-surface daily and synoptic temperature variability over North America compared to
405 Eurasia. Our analysis combines theoretical arguments, simulations with two idealized GCMs and
406 simulations with a comprehensive climate model – GFDL CM2.5-FLOR – in which the Rockies
407 and the Asian orography are separately flattened. These allow us to quantify the impacts these

408 mountain ranges have on temperature variability over North America and Eurasia, and suggest
409 that large-scale Northern Hemisphere orography is responsible for roughly 25% of the difference
410 in variability.

411 Large-scale orography enhances downstream temperature variability by meridionally compress-
412 ing downstream isentropes and reduces upstream temperature variability because upstream isen-
413 tropes are pulled apart. At the same time, the preferential deflection of the flow towards the pole-
414 ward flank of the orography, together with the presence of high latitude easterlies, can cause this
415 pattern to be reversed at high latitudes, with enhanced variance on the northwest flank and reduced
416 variance on the northeast flank of the orography (in the Northern Hemisphere). We have also
417 shown that the orography's aspect ratio can cause substantial differences in the pattern of variabil-
418 ity; for instance, a meridional ridge, resembling the Rockies, induces a stronger local enhancement
419 of temperature variance on its downstream flank, whereas for circular orography the enhanced
420 variance is further downstream. Finally, latent heat anomalies reinforce temperature anomalies
421 created by advection, as anomalously warm air originating from low latitudes condenses water as
422 it moves poleward and rises.

423 Most of North America is downstream of the Rockies, so wintertime temperature variability
424 is enhanced there, while the Asian orography is on the eastern edge of Eurasia, so temperature
425 variability is damped over most of Eurasia. An important exception is the southeast Asian littoral,
426 which is east of the orography but exhibits reduced temperature variability due to the Asian moun-
427 tains. This is partly because these regions are at latitudes of mean easterlies, or in the transition
428 from mean westerlies to mean easterlies, and hence are upstream of the Himalayas. Another fac-
429 tor is interference by the Asian orography with the energization of eddies over the Asian continent
430 and the Pacific storm track by downstream development. This results in weaker winter storms
431 and reduced variability over the east Asian coast, the Pacific and North America. The reduction

432 in variability over North America due to the presence of the Asian orography is approximately as
433 large as the increase due to the presence of the Rockies.

434 Orography also enhances downstream skewness, as regions to the north of the enhanced tem-
435 perature gradient have more positively skewed temperatures and regions to the south have more
436 negatively skewed temperatures. In the FLOR simulations, the Himalayas and the Tibetan Plateau
437 are found to increase temperature skewness in the northwest Pacific by about 30%.

438 The remaining difference in synoptic temperature variability over North America compared to
439 Eurasia is primarily due to a combination of diabatic heating in the Pacific warm pool region, air-
440 sea fluxes over the warm Kuroshio current and eddy sensible heat flux convergence in the Pacific
441 storm track (Valdes and Hoskins (1989); Held et al. (2002)). The smaller width of the North
442 American continent and its northwest-southeast sloping western coastline may also be important
443 – Brayshaw et al. (2009) explored how this influences the North Atlantic storm track. Separating
444 out these different factors, and the non-linear interactions between them, is an important next step.

445 The dominant control of horizontal advection on winter synoptic temperature variability is a
446 powerful tool for understanding the regional pattern of temperature variability, in today's climate
447 and how it may change in the future. This simplifies the problem to understanding the boreal winter
448 stationary wave pattern, for which there is a large body of literature that can be drawn on (e.g.,
449 Hoskins and Karoly (1981); Held (1983); Held et al. (2002)), though differences in mixing lengths,
450 for instance due to orographic interference with downstream development, are an important caveat.
451 Similarly, past and future changes in temperature variability can potentially be tied to changes in
452 the stationary wave pattern (see e.g., Löfverström et al. (2014) and Simpson et al. (2016) for
453 investigations of past and future changes in Northern Hemisphere stationary waves). More work
454 is needed to better understand the impact of orography on mixing lengths, as well as to account
455 for land surface processes such as soil-moisture, which affect temperature variability, particularly

456 during summer. These factors are also important for temperature extremes, particularly over land,
457 where winter temperatures at almost all latitudes are negatively skewed. Nevertheless, the basic
458 dynamics we describe here are robustly seen in idealized GCMs and in comprehensive climate
459 models, and provide an important first step in explaining why North America experiences more
460 wintertime temperature variability than Eurasia.

461 *Acknowledgment.* We thank Daniel Koll, Rodrigo Caballero and Gabriel Vecchi for helpful con-
462 versations over the course of this work. The manuscript was much improved by comments from
463 three anonymous reviewers, as well as feedback from Daniel Koll and Paul O’Gorman on earlier
464 drafts. Gabriel Vecchi kindly provided the computational resources to perform the FLOR simu-
465 lations. This work was partly supported by NSF grant AGS-1623218, “Collaborative Research:
466 Using a Hierarchy of Models to Constrain the Temperature Dependence of Climate Sensitivity”.

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583 TABLE 1. Variance of December-January-February (DJF) 850hPa synoptic temperature over Eurasia (40°E -
 584 120°E and 30°N - 75°N) and North America (240°E - 280°E and 30°N - 75°N) in the FLOR simulations and ob-
 585 served variances from 1979-2012. All units are K^2 and the plus/minus values show the standard deviations of
 586 the interannual variability.

	MERRA reanalysis 1979-2012	Control	no-Tibet	no-Rockies
North America	15.7 ± 2.6	18.1 ± 2.4	19.4 ± 2.7	16.8 ± 2.9
Eurasia	8.6 ± 2.0	12.6 ± 2.3	14.0 ± 2.8	12.4 ± 2.2
Difference	7.1	5.5	5.4	4.4

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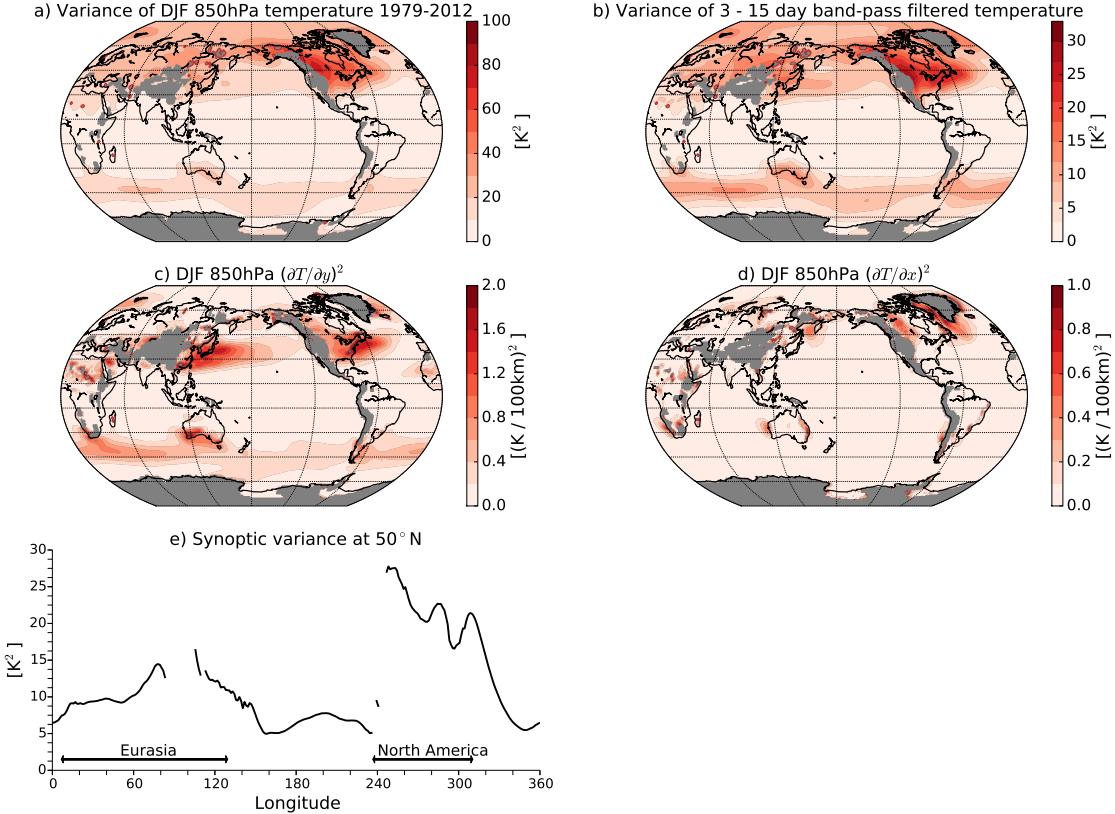
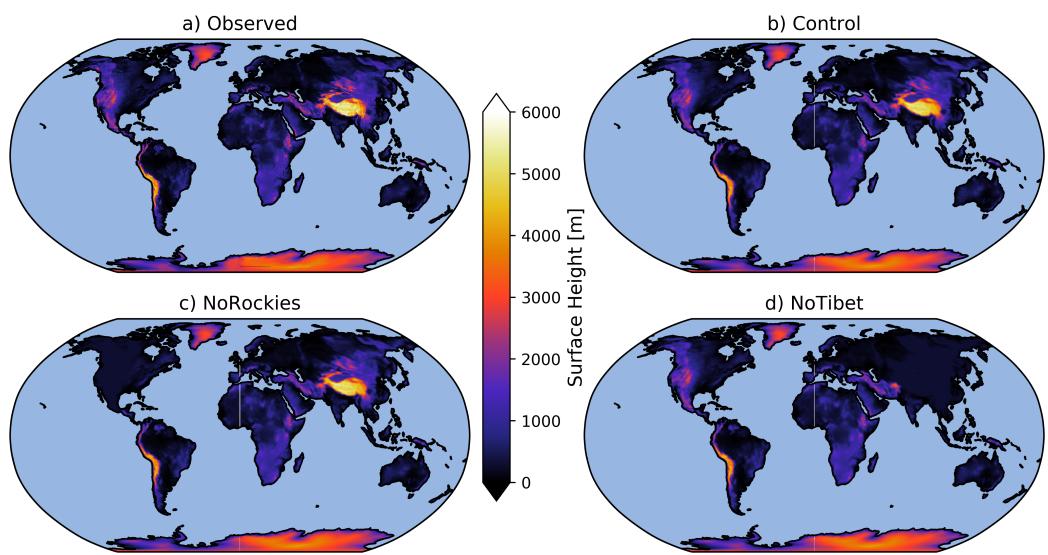


FIG. 1. a) Variance of December-January-February (DJF) 850hPa daily temperature for the period 1979 to 659 2012, calculated using data taken from the MERRA reanalysis dataset. Locations where topography intrudes
660 through 850hPa are masked in gray. b) Same as panel a, but the data are filtered using a fourth-order Butterworth
661 filter to only retain power at synoptic time-scales, here defined as 3 to 15 days. c) Climatological DJF squared
662 meridional temperature gradients for the same data. d) Climatological DJF squared zonal temperature gradients
663 for the same data. e) Profile of synoptic-scale 850hPa temperature variance at 50°N. Gaps in the profiles show
664 where topography intrudes into the 850hPa level.
665



666 FIG. 2. a) Observed topography of Earth, taken from the ETOPO5 dataset, with 5 minute resolution. b)
 667 Topography in the control simulation of FLOR. c) Topography in the no-Rockies simulation. d) Topography in
 668 the no-Tibet simulation.

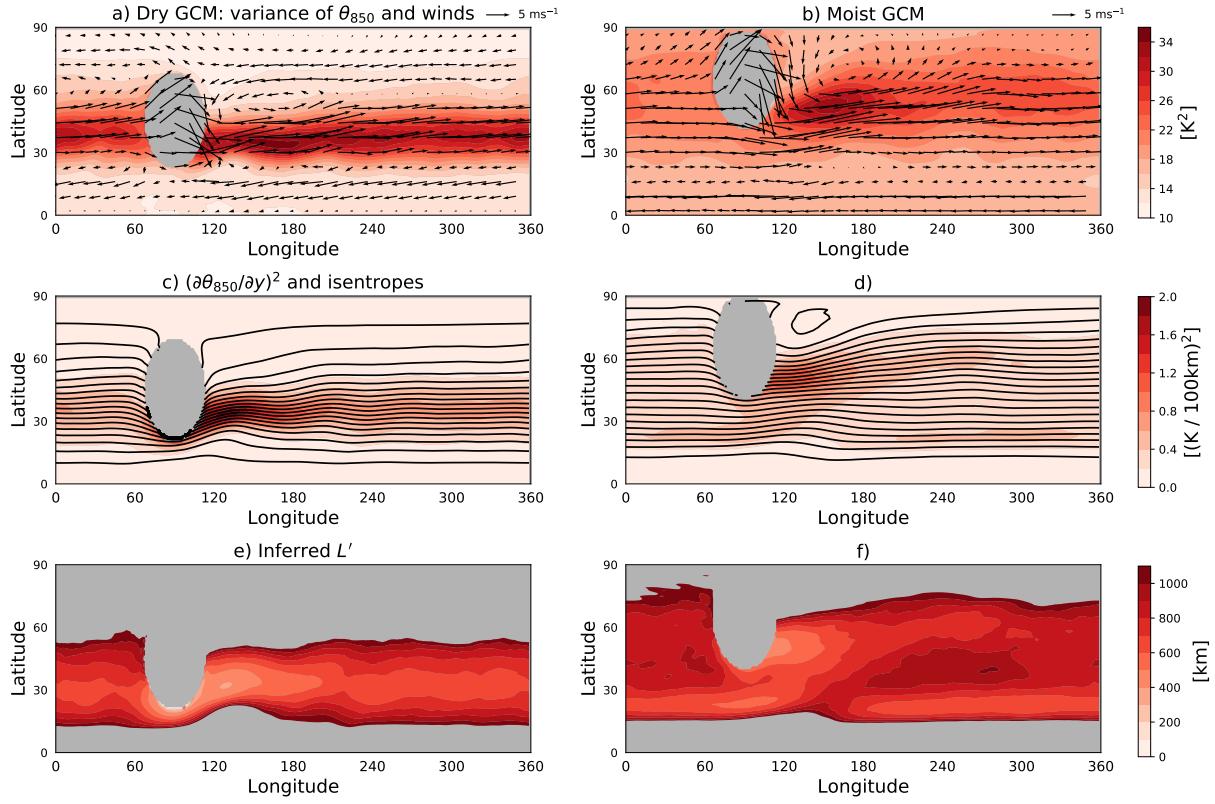


FIG. 3. a) Variance of synoptic (3 to 15 days) 850hPa potential temperature (contours) and total wind vectors (arrows) in a simulation with the dry GCM and maximum orographic height of 4km. b) Same as panel a) but for the simulation with the moist GCM with 4km orography. c) Squared meridional gradient of time-averaged potential temperature (colored contours) and isentropes (black contours, with contour interval 2K) at 850hPa for the same simulation as in panel a). d) Same as panel c) but for the simulation with the moist GCM with 4km orography. e) Inferred mixing length L' for the simulation with the dry GCM and maximum orographic height of 4km. f) Same as panel d) but for the simulation with the moist GCM with 4km orography. In all panels gray indicates locations with surface pressure less than 850hPa or, in the bottom panels, where values are outside the colorbar range. In a) and b) the winds are taken from the 0.85- σ level so that the flow over and around the orography is visible.

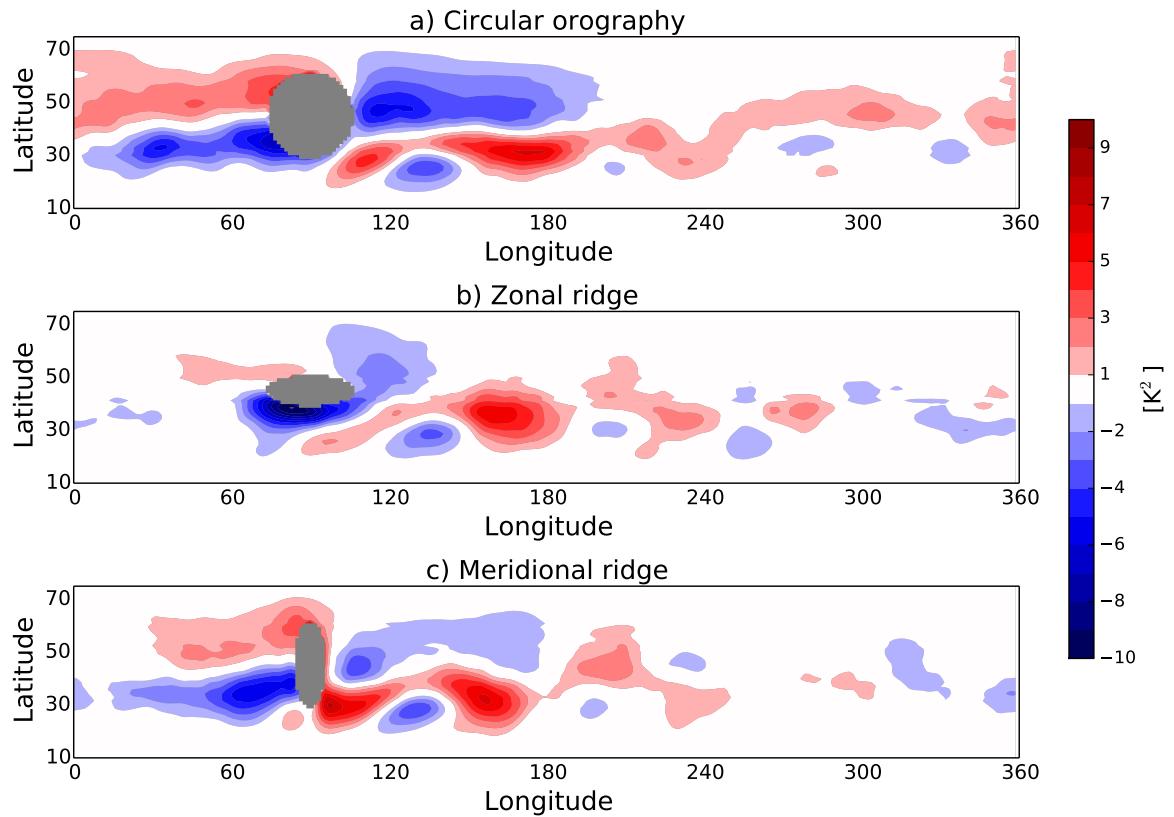


FIG. 4. a), b), c) Zonal anomalies in the variance of synoptic (3-15 day) 850hPa potential temperature in the dry GCM simulations with $H = 4\text{km}$ and the circular Gaussian orography (a, $\alpha = \beta = 15^\circ$), the zonal ridge (b, $\alpha = 5^\circ$ and $\beta = 15^\circ$) and the meridional ridge (c, $\alpha = 15^\circ$ and $\beta = 5^\circ$).

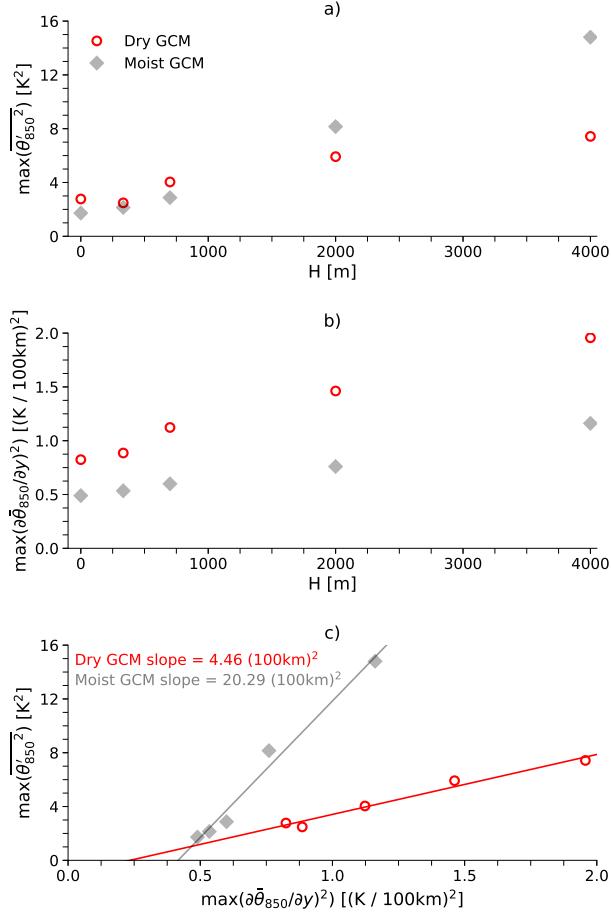


FIG. 5. a) Maximum anomalous 850hPa potential temperature variance ($\max(\overline{\theta'_{850}^2})$) as a function of mountain height, H , in the simulations with the dry GCM (red circles) and with the moist GCM (gray diamonds).
 b) Maximum zonal anomaly of the squared meridional potential temperature gradient at 850hPa ($\max((\partial\bar{\theta}_{850}/\partial y)^2)$) as a function of H in the simulations with the idealized GCMs.
 c) $\max(\overline{\theta'_{850}^2})$ versus $\max((\partial\bar{\theta}_{850}/\partial y)^2)$ in the simulations with the idealized GCMs. The lines show linear least-squares fits to the two sets of simulations.

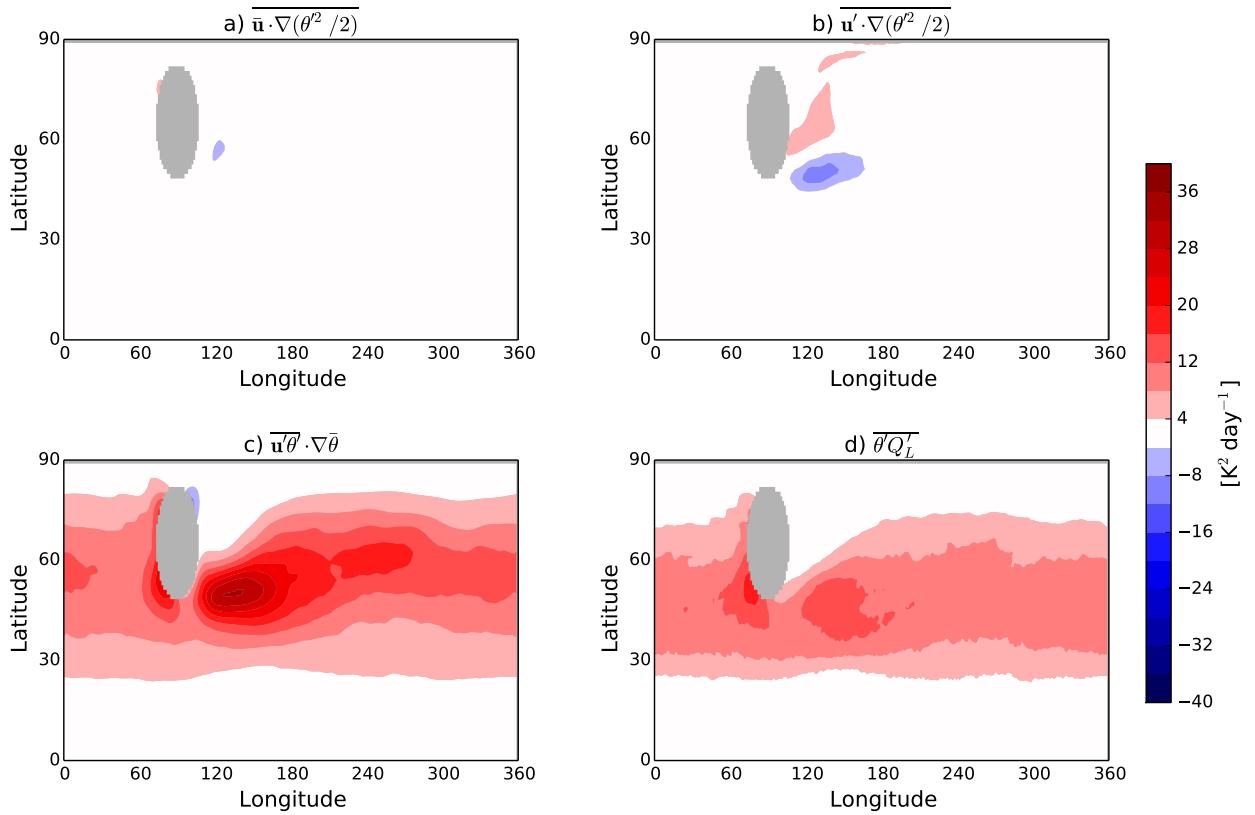
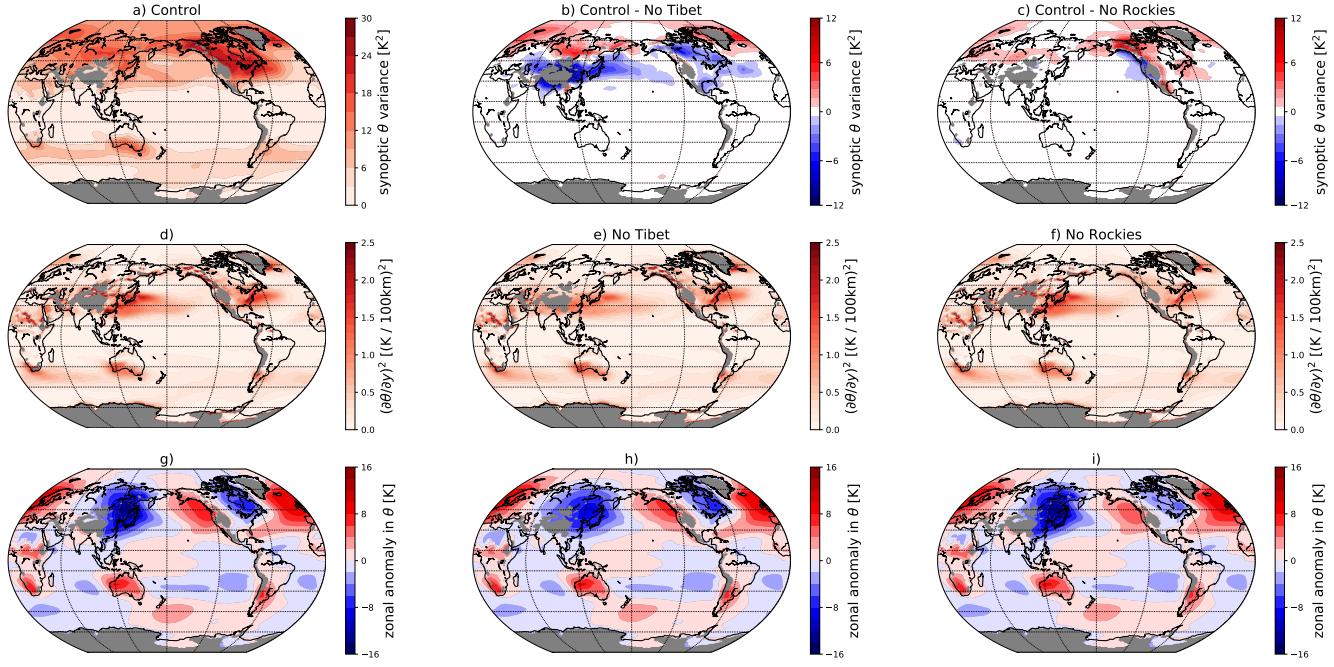
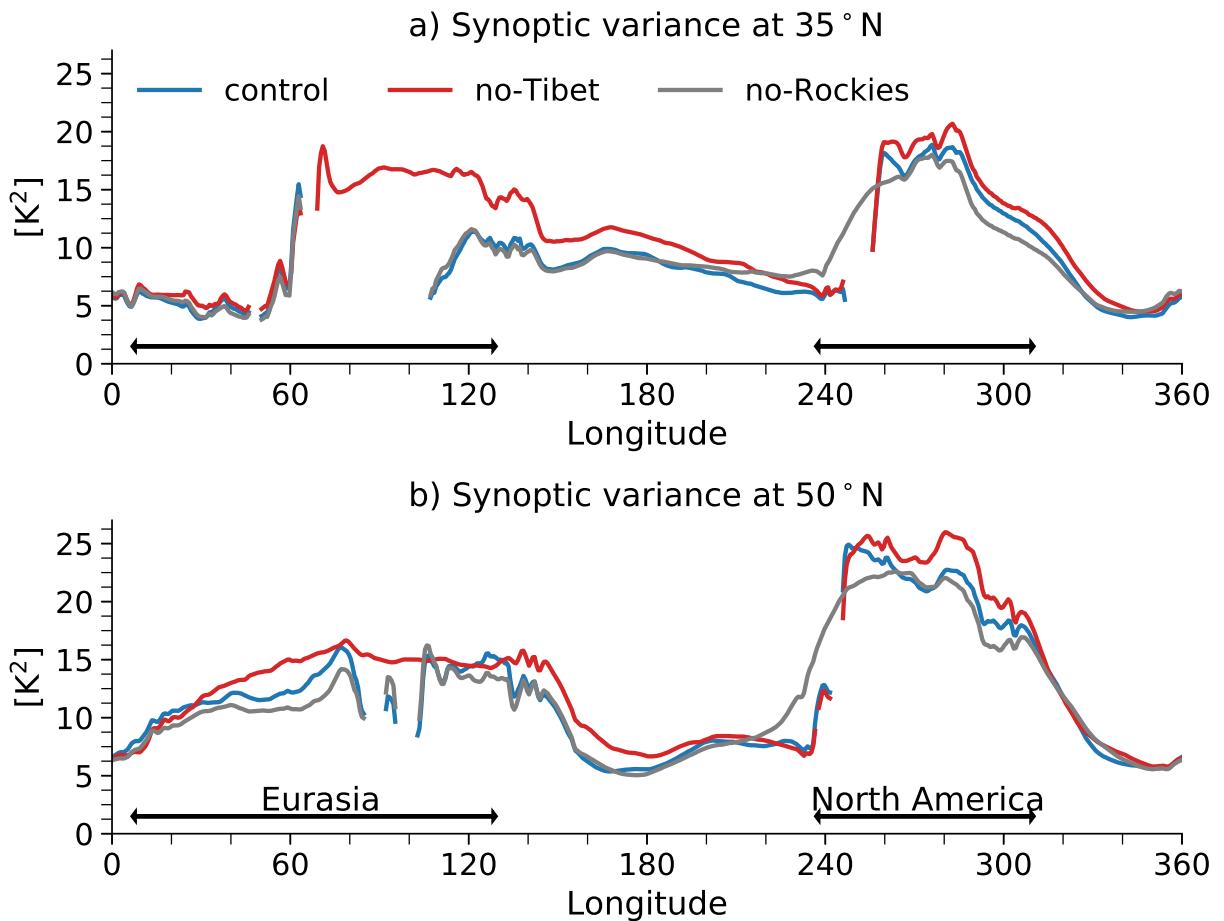


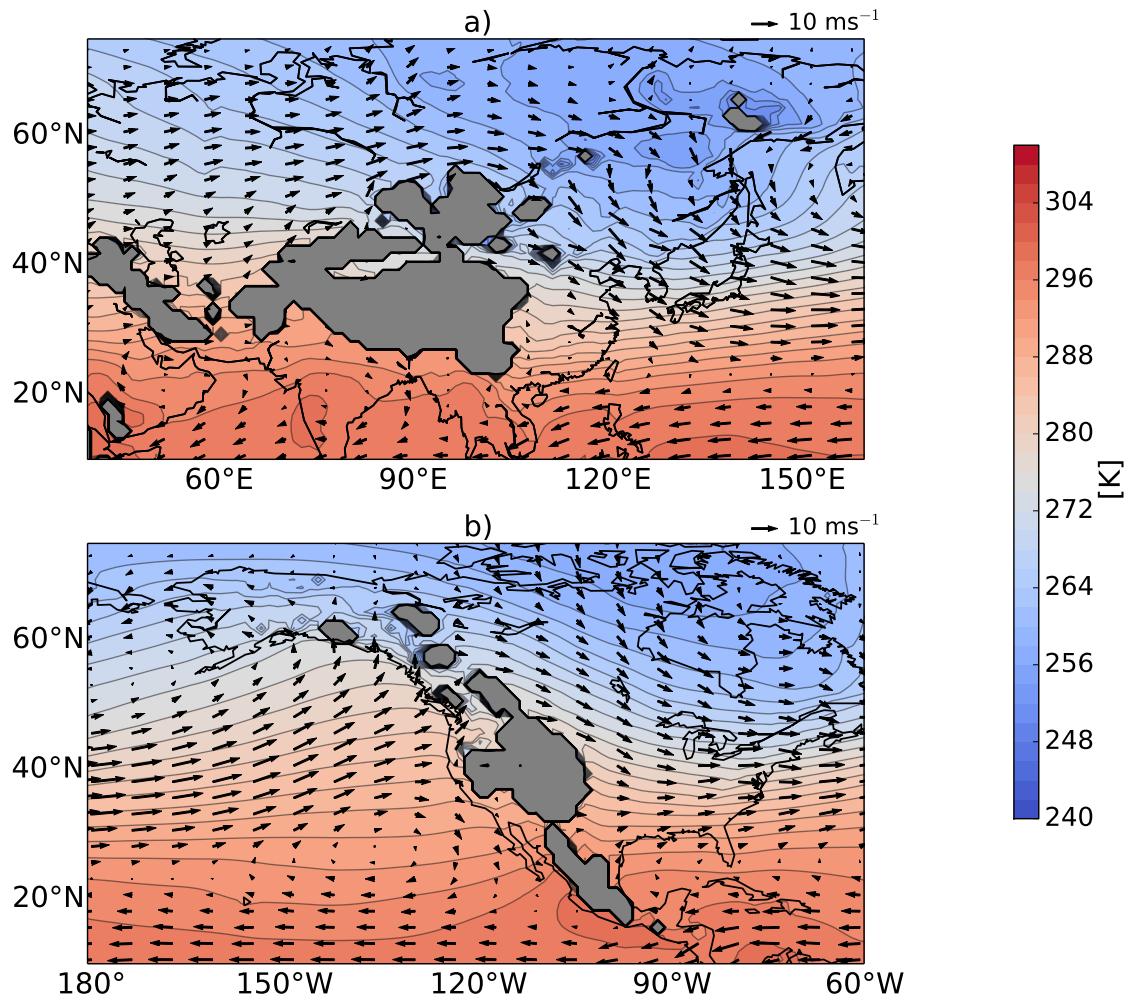
FIG. 6. Panels a), b) and c): advective terms in the 850hPa potential temperature variance budget from a simulation with the moist GCM and a mountain height of 4km. Locations where topography intrudes through 850hPa are masked in gray. Panel d): the contribution of latent heating fluctuations to 850hPa potential temperature variance in the same simulation.



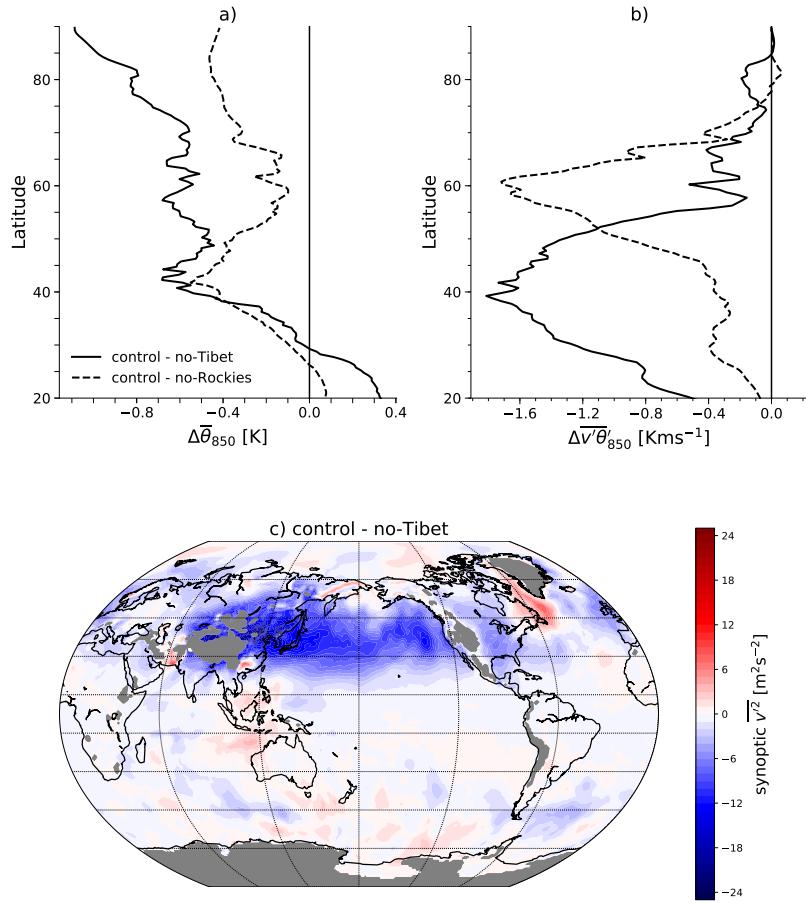
694 FIG. 7. a) Synoptic-scale variance of DJF 850hPa potential temperature in the control simulation with the
 695 comprehensive climate model, FLOR. b) Difference in synoptic-scale variance between the control simulation
 696 and the no-Tibet simulation. c) Difference in synoptic-scale variance between the control simulation and the
 697 no-Rockies simulation. d) DJF squared meridional potential temperature gradients in the control simulation. e)
 698 DJF squared meridional potential temperature gradients in the no-Tibet simulation. f) DJF squared meridional
 699 potential temperature gradients in the no-Rockies simulation. g), h), i) DJF zonal anomalies in 850hPa potential
 700 temperature in the same simulations. Locations where topography intrudes through 850hPa are masked in gray.



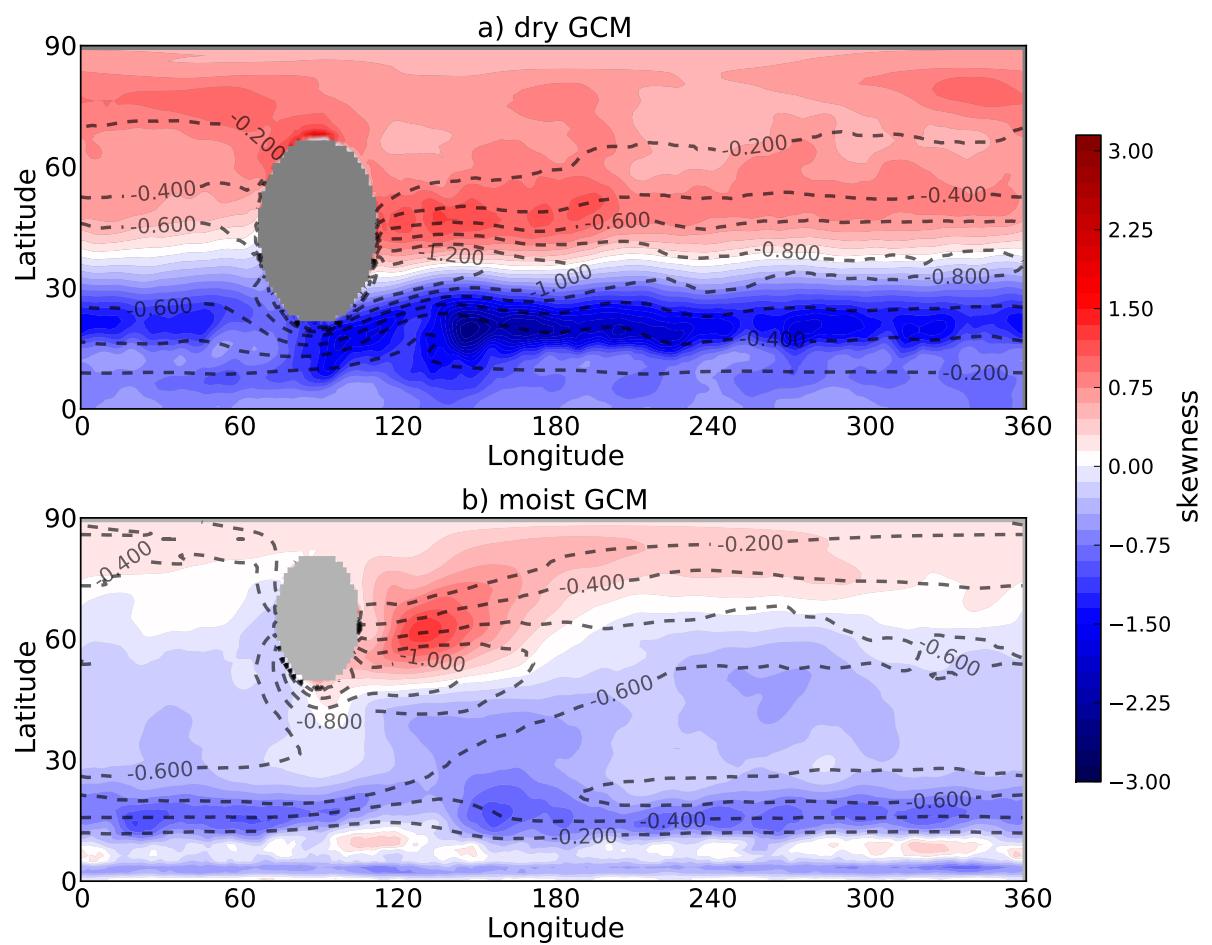
701 FIG. 8. a) Profiles taken at 35°N of synoptic-scale variance of DJF 850hPa potential temperature in the three
 702 simulations with FLOR. b) Profiles taken at 50°N. Gaps in the profiles show where topography intrudes into the
 703 850hPa level.



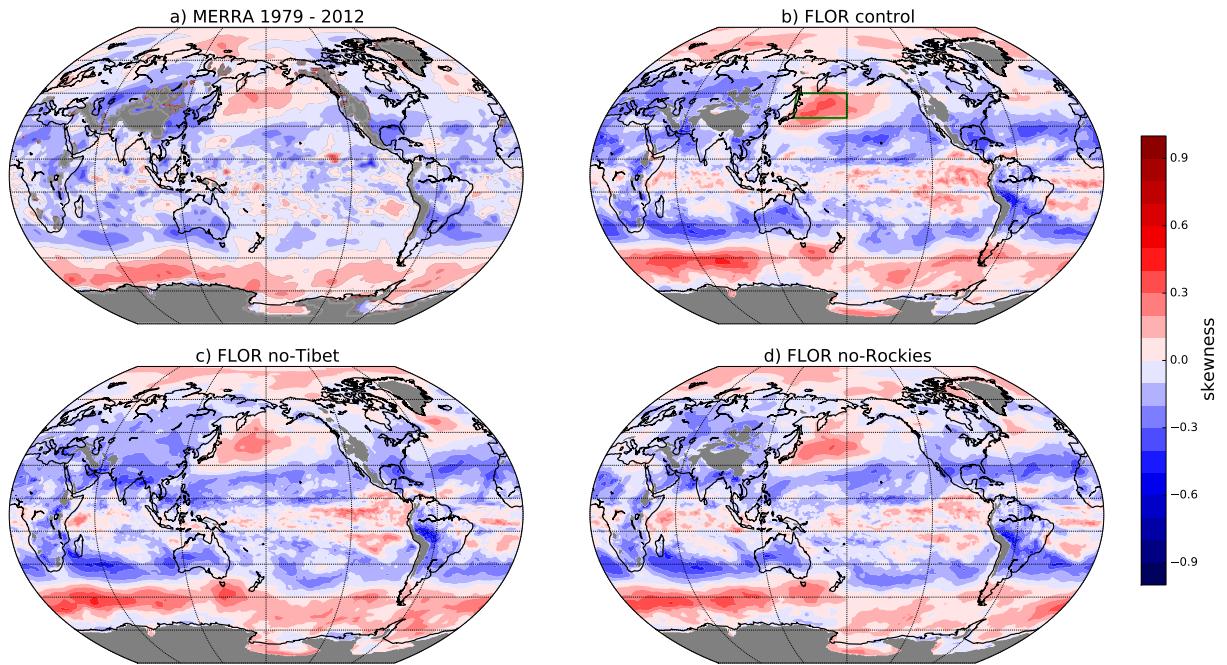
704 FIG. 9. a) DJF 850hPa temperature (contours) and total wind vectors in the vicinity of the Tibetan Plateau,
 705 averaged over the period 1979 to 2012. Data are taken from the MERRA reanalysis dataset. Locations where
 706 topography intrudes through 850hPa are masked in gray. b) Same for the region near the Rocky Mountains.



707 FIG. 10. a) Difference in zonal-mean θ_{850} between the control simulation with FLOR and the no-Tibet
 708 simulation (solid line) and difference between the control simulation and the no-Rockies simulation (dashed
 709 line). b) Differences in transient eddy potential temperature flux in the same simulations. c) Difference in DJF
 710 synoptic 850hPa eddy kinetic energy (v'^2) between the control simulation and the no-Tibet simulation.



711 FIG. 11. a) Skewness of 850hPa synoptic temperatures (colored contours) and 850hPa meridional temperature
 712 gradients (black contours, contour interval = $0.2\text{K}(100\text{km})^{-1}$) in the dry GCM simulation with $H = 4\text{km}$. b)
 713 Same for the simulation with the moist GCM. The meridional gradient contour interval is $0.2\text{K}/100\text{km}$ in both
 714 panels.



715 FIG. 12. a) Skewness of DJF 850hPa synoptic temperatures for the period 1979-2012 in the MERRA data.
 716 b) Skewness of DJF 850hPa synoptic temperatures in the control simulation with FLOR. c) Skewness of DJF
 717 850hPa synoptic temperatures in the no-Tibet simulation with FLOR. d) Skewness of DJF 850hPa synoptic
 718 temperatures in the no-Rockies simulation with FLOR.