

1 **Improved parameterization of snow albedo in WRF + Noah. Part II:**

2 **Applicability to snow estimates for the Tibetan Plateau**

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18

19 **Abstract**

20 Snow albedo is important to the land surface energy balance and to the water cycle.
21 During snowfall and subsequent snowmelt, snow albedo is usually parameterized as
22 functions of snow related variables in land surface models. However, the default snow
23 albedo scheme in the widely used Noah land surface model shows evident shortcomings
24 in land-atmosphere interactions estimates during snow events on the Tibetan Plateau.
25 Here, we demonstrate that our improved snow albedo scheme performs well after
26 including snow depth as an additional factor. By coupling the WRF and Noah models,
27 this study comprehensively evaluates the performance of the improved snow albedo
28 scheme in simulating eight snow events on the Tibetan Plateau. The modeling results
29 are compared with WRF run with the default Noah scheme and in situ observations.
30 The improved snow albedo scheme significantly outperforms the default Noah scheme
31 in relation to air temperature, albedo and sensible heat flux estimates, by alleviating
32 cold bias estimates, albedo overestimates and sensible heat flux underestimates,
33 respectively. This in turn contributes to more accurate reproductions of snow event
34 evolution. The averaged RMSE relative reductions (and relative increase in correlation
35 coefficients) for air temperature, albedo, sensible heat flux and snow depth reach 27 %
36 (5 %), 32 % (69 %), 13 % (17 %) and 21 % (108 %) respectively. These results
37 demonstrate the strong potential of our improved snow albedo parameterization scheme
38 for snow event simulations on the Tibetan Plateau. Our study provides a theoretical
39 reference for researchers committed to further improving the snow albedo
40 parameterization scheme.

41 **Keywords:** WRF; snow albedo parameterization; turbulent heat and vapor fluxes;
42 Tibetan Plateau

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44 **1 Introduction**

45 The surface albedo directly determines the proportion of incident solar radiation that is
46 absorbed by the surface, and is an important parameter in climate and land surface
47 models (LSMs) (Sellers et al., 1996). Small changes in surface albedo can affect the
48 energy balance in the land-atmosphere system, and can drive both local and global
49 climate change (Bloch, 1964).

50 Surface albedo changes dramatically during snowfall and snowmelt cycles. Much
51 research has been carried out to identify the factors that influence these changes,
52 including the effects of terrain shielding, altitude, sky conditions, vegetation, and snow
53 properties such as grain size, liquid water content, depth, and impurities (Warren and
54 Wiscombe, 1980; Wiscombe and Warren, 1980; Aoki et al., 2003; Jonsell et al., 2003;
55 Hansen and Nazarenko, 2004; Liang et al., 2005; Wang et al., 2015; He et al., 2018a).
56 This body of research has led to the development of many parameterization schemes
57 for surface albedo (Oerlemans and Knap, 1998; Wang et al., 2007; Bao et al., 2008; Li
58 and Hu, 2009; Gardner and Sharp, 2010; Kuipers Munneke et al., 2011; Malik et al.,
59 2014; Dang et al., 2015; He et al., 2017, 2018b; Meng and Li, 2019; Saito et al., 2019;
60 Wang et al., 2020). Most snow albedo parameterization schemes depend on statistical
61 empirical formulas and constant parameters, rather than representing physical snow-
62 albedo feedback processes. To improve the performance of snow albedo
63 parameterization schemes for simulating land-atmosphere interactions, Bao and Lyu
64 (2009) added consideration of solar zenith angle to a regional climate model, which
65 resulted in a 1.2 °C temperature increase, and considerably improved the cold bias in
66 East Asia and improved the representation of diurnal ground temperature changes in
67 northwest China. Park and Park (2016) investigated the effect of vegetation on snow
68 covered surface albedo and improved the winter surface albedo estimates from their
69 LSM by including leaf and stem indices in the snow albedo parameterization scheme,
70 which reduced the root mean square error (RMSE) by 69 %. Zhong et al. (2017)
71 considered aerosol radiative effects on snow processes in their simulations and

72 successfully reproduced the snow albedo and snow depth. Fresh snow albedo depends
73 on snow depth, and albedo parameterization schemes that fail to account for this
74 generally overestimate the snow depth. To address this, Wang et al. (2020) developed a
75 new albedo scheme for fresh snow, which accounts for the relationship between fresh
76 snow albedo, snow grain size and snow depth, resulting in improved snow depth
77 estimates during the snow ablation period on the Tibetan Plateau. This highlights the
78 importance of accounting for the effect of snow depth on fresh snow albedo.

79 A coupled land-atmosphere model can provide useful insights into conditions on the
80 Tibetan Plateau, where the terrain is complex and there are few, and unevenly
81 distributed observation stations (Maussion et al., 2011; Yuan et al., 2016; Norris et al.,
82 2017; Bonekamp et al., 2018; Rahimi et al., 2019). However, the parameterization
83 scheme for surface albedo in the Noah LSM, which is currently the most widely used
84 LSM, does not account for all the factors that influence albedo. It includes many
85 predetermined parameters and an approximate treatment of vegetation, soil and snow,
86 which can result in some inaccuracies in the estimated surface albedo (Wen et al., 2011;
87 Liu et al., 2019). For example, the surface albedo parameterization scheme in the Noah
88 LSM considers snow cover and age, but ignores other snow related factors, such as
89 snow depth, that can drive dramatic changes in albedo (Ek et al., 2003). This makes it
90 inappropriate to use the Noah LSM to characterize changes in snow albedo that follow
91 from snowfall and melt processes in complex topographic areas. However, the Noah
92 LSM appears to be the most readily available snow albedo scheme for long term climate
93 modeling research (Rai et al., 2019). Despite its shortcomings, the Noah albedo
94 parameterization scheme does provide substantial improvements to estimates of the
95 magnitude and timing of both the peak snowfall amount and the maximum snow cover
96 extent, following from the scheme's consideration of snow albedo decay and liquid
97 water refreezing (Livneh et al., 2010). The above issues represent opportunities for
98 improvements to be made to the snow albedo parameterization scheme in the Noah
99 LSM.

100 The use of an advanced snow albedo parameterization scheme is crucial for accurate
101 estimation of land-atmosphere interactions over the Tibetan Plateau, where the snow-
102 albedo effect is extremely strong. It has been shown that the Weather Research and
103 Forecasting (WRF; Skamarock et al., 2008) model, when coupled with the default Noah
104 albedo parameterization scheme, results in an apparent cold bias over the Tibetan
105 Plateau (Gao et al., 2015; Meng et al., 2018; Liu et al., 2019). This bias can be reduced
106 by including albedo products from the Moderate Resolution Imaging
107 Spectroradiometer (MODIS) and an additional snow depth parameter as independent
108 variables in the Noah albedo parameterization scheme (Liu, 2020). This approach is not
109 the same as assimilating satellite retrieved snow related products into the LSM, which
110 has also been shown to lead to improvements (Xu and Shu, 2014; Zhang et al., 2014;
111 Lin et al., 2016; Xue et al., 2019). This improved snow albedo scheme has been
112 successfully implemented in the WRF model, coupled with Noah, to simulate land-
113 atmosphere interactions during a regional heavy snow event on the Tibetan Plateau (Liu,
114 2020). However, it has not been shown that the improvements that follow from the
115 improved snow albedo scheme are universal over the Tibetan Plateau, and this should
116 be studied further. Severe snowfall occurs often over the southern Tibetan Plateau,
117 while snowfall over the eastern Tibetan Plateau is generally of relatively weak intensity,
118 and the rate of snowmelt varies widely depending on the heterogeneous underlying
119 surfaces. This makes it necessary to carry out numerical experiments that focus on snow
120 events over the eastern and southern Tibetan Plateau to assess how reliably the
121 improved scheme can be used to characterize different snowfall intensities and
122 snowmelt processes.

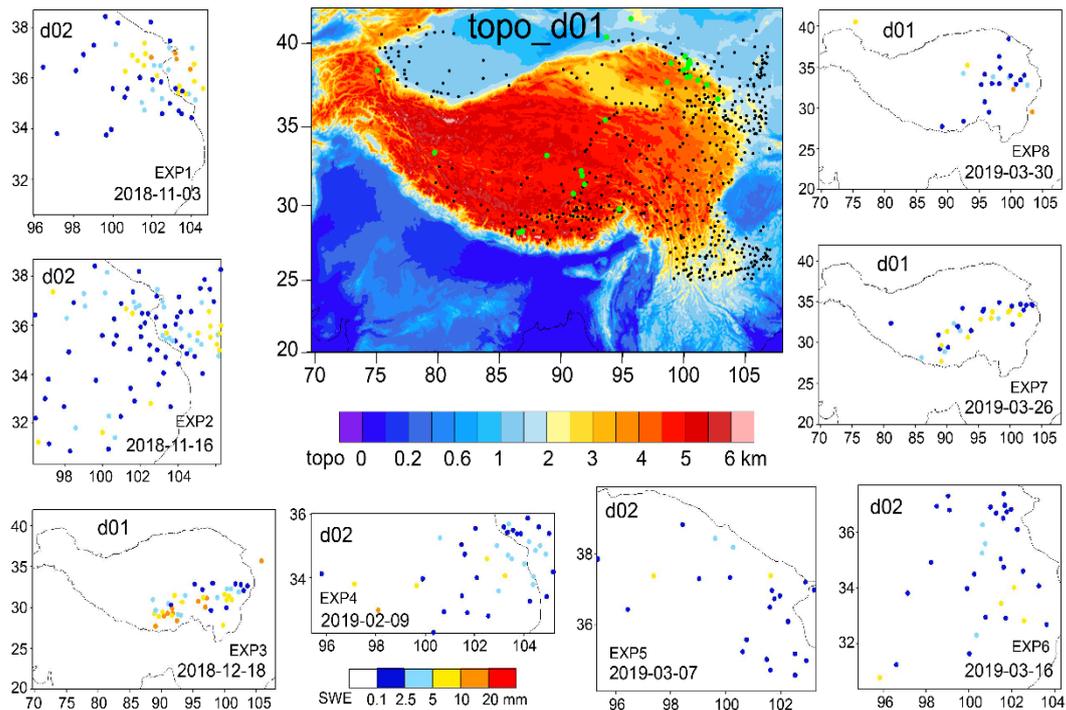
123 In this study, we selected eight moderate to snowstorm events that occurred over the
124 southern and eastern Tibetan Plateau to assess the universality of the improvements
125 offered by our improved snow albedo scheme in WRF coupled with the Noah LSM.
126 For each snow event, two numerical experiments were carried out: one implementing
127 the default Noah snow albedo scheme, and the other implementing our improved snow
128 albedo scheme. The model performance was assessed through comparison of the

129 modeled air temperature, albedo, snow depth, turbulent heat and vapor fluxes with
130 ground observations. The aim of this study is to explore the potential of our improved
131 snow albedo parameterization scheme to simulate snow events over the whole Tibetan
132 Plateau more accurately than can be done using the standard default scheme. We hope
133 that this study will also provide a useful reference for researchers working to develop
134 and improve this, and other albedo parameterization schemes.

135 **2 Data and methodology**

136 **2.1 Description of snow events**

137 We selected China Meteorological Administration (CMA) national observation stations
138 on the Tibetan Plateau, and in the surrounding regions, with elevations exceeding 1000
139 m. This resulted in a total of 502 stations (Figure 1 shows their distribution). Snowfall
140 events were identified from the hourly air temperature and precipitation observations
141 from all 502 stations when the air temperature was below 0 °C. Daily snowfall amounts
142 were calculated, using 08:00 Beijing Standard Time (BST) as the start and end time for
143 each day. The standards to define snowfall grade were taken from the China
144 Meteorological Standardization Network (<http://www.cmastd.cn/>). Using these
145 standards, eight different grades of snowfall event were considered in our study,
146 including moderate, heavy and snowstorm. Most of the snowfall events took place over
147 the eastern Tibetan Plateau, and some occurred in a large region across the southern and
148 central to the eastern Tibetan Plateau. The maximum daily snowfall amount from all
149 snowfall events exceeded 8 mm, and four events resulted in more than 10 mm daily
150 snowfall, making these snowstorm grade events. Snow depth is much greater on the
151 southern Tibetan Plateau (> 50 cm) than it is on the eastern and central Tibetan Plateau
152 (<=20 cm). The description of eight snowfall events, including the date and location of
153 moderate to snowstorm grade events, the maximum snow depth and daily snowfall
154 amount, are detailed in Table 1. The daily snowfall amounts from the eight snowfall
155 events are shown in Figure 1.



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157 **Figure 1.** WRF domains of d01 and d02, and CMA observations of daily snowfall
 158 amount in color solid circles for the eight experiments. The topographical height of d01
 159 is shaded with black solid circles indicating the locations of the CMA stations, and
 160 green solid circles indicating the locations of the CAS stations and the Qilian Mountains
 161 integrated observatory network.

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170 **Table 1** Description of eight snowfall events using China Meteorological
 171 Administration observations.

Snow events	Event 1	Event 2	Event 3	Event 4	Event 5	Event 6	Event 7	Event 8
Date of moderate to snowstorm	Nov. 3-6, 2018	Nov. 15-17, 2018	Dec. 17-18, 2018	Feb. 9, 2019	Mar. 7-8, 2019	Mar. 16, 2019	Mar. 26, 2019	Mar. 30-Apr. 1, 2019
Region in the Tibetan Plateau	eastern	eastern	southern, central, eastern	eastern	eastern	eastern	southern, central, eastern	central, eastern
Maximum daily snowfall amount (unit: mm)	13.6	9.2	18.7	11	9.3	9.1	8.7	14.3
Maximum snow depth (unit: cm)	18	16	65	20	13	9	53	18

172 **2.2 Model description and experiment configuration**

173 The WRF model (Skamarock et al., 2008), version 3.7.1, coupled with the Noah LSM,
 174 was used to simulate the eight snowfall events in this study. It is a fully compressible,
 175 non-hydrostatic model and includes a run time hydrostatic option. Vertical levels are
 176 determined using a mass based terrain following hydrostatic pressure coordinate, and
 177 calculations are performed on an Arakawa C grid. The model uses 2nd and 3rd order
 178 Runge-Kutta time integration schemes, and 2nd to 6th order advection schemes in both
 179 the horizontal and vertical direction.

180 The extremely steep terrain on the central and southern Tibetan Plateau led to model
 181 instability and failure for snowfall events 3, 7 and 8 when a relatively fine horizontal
 182 resolution of 1 km was used; however, the calculations remained stable when the
 183 resolution was increased to 5 km. *We therefore used two ways nested modeling domains*

184 in our model configuration for snowfall events 1, 2, 4, 5 and 6, and a single modeling
185 domain for snowfall events 3, 7 and 8. The coarse domain (d01) was used to simulate
186 synoptic scale atmospheric conditions over 20.0-42.0° N and 69.7-108.0° E with a
187 horizontal resolution of 5 km. The inner domain (d02) had a horizontal resolution of 1
188 km, and event 1 occupied 876×966 grid cells, event 2 occupied 976×1001 cells, event
189 4 was resolved by 966×451 cells, event 5 was calculated over 781×686 cells, and
190 event 6 covered 926×881 cells. The vertical structure of both domains included 35
191 unevenly spaced layers and extended up to 50 hPa. The model was configured to use
192 the Noah LSM in d01 and d02 to describe all land-atmosphere interactions; the
193 Thompson scheme to represent microphysical processes; the Dudhia scheme to
194 represent shortwave radiation, the RRTM scheme to describe longwave radiation; the
195 YSU scheme to describe the planetary boundary layer; and only in d01 to use the Kain-
196 Fritsch cumulus parameterization scheme for clouds.

197 We conducted numerical experiments to simulate snow event 1 (EXP1), event 2 (EXP2),
198 event 3 (EXP3), event 4 (EXP4), event 5 (EXP5), event 6 (EXP6), event 7 (EXP7) and
199 event 8 (EXP8). The model domains for the experiments are shown in Figure 1. Each
200 experiment included two model simulations: one implementing the default Noah snow
201 albedo parameterization scheme, and the other implementing a new improved snow
202 albedo scheme. *The new improved snow albedo was developed based on conversion*
203 *formula from MODIS narrowband spectral reflectance to broadband albedo following*
204 *Liang (2000) and albedo calculation formula about fresh snow, firn and bare ground*
205 *albedo, snow age and depth following Oerlemans and Knap (1998). MODIS broadband*
206 *albedo and WRF modeled snow depth and age were used to estimate the related*
207 *parameters i.e., firn albedo and scales of snow depth and age through nonlinear fitting*
208 *of the above albedo calculation formula. The final nonlinear fitting results produced*
209 *the new improved snow albedo scheme, seeing the equations (3) and (4). The snow*
210 *albedo is parameterized using Eq. (1) and (2) in the default Noah land surface scheme*
211 *(Livneh et al., 2010), and using Eq. (3) and (4) in the improved scheme (Liu, 2020):*

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$$\alpha_{\text{snow}} = \alpha_{\text{max}} \times A^{t^B} \quad (1)$$

$$\alpha = \alpha_{bg} + sc \times (\alpha_{\text{snow}} - \alpha_{bg}) \quad (2)$$

$$\alpha_{\text{snow}} = 0.13 + 0.66e^{\left(\frac{t}{1.38}\right)} \quad (3)$$

$$\alpha = \alpha_{\text{snow}} + (0.19 - \alpha_{\text{snow}})e^{\left(\frac{-d}{0.11}\right)} \quad (4)$$

215 where A and B are constants, equal to 0.94 and 0.58, respectively, for snow
216 accumulation periods, and are 0.82 and 0.46, respectively, for other periods; α_{bg} is the
217 background albedo, which depends on the land cover type; sc is snow cover, and
218 ranges from 0 to 1; α_{max} is fresh snow albedo; α_{snow} is snow albedo; t is the snow
219 age in units of days; d is snow depth in meters.

220 The fifth generation European Centre for Medium Range Weather Forecasts (ECMWF)
221 reanalysis dataset (ERA5), with a spatial resolution of $0.25^\circ \times 0.25^\circ$ and 3 h temporal
222 resolution, provided initial and boundary conditions for our numerical experiments. The
223 ERA data were calculated using 4DVar data assimilation in CY41R2 from ECMWF's
224 Integrated Forecast System, with 137 vertical hybrid sigma/pressure levels, extending
225 to 0.01 hPa. ERA5 is freely available from the website
226 <https://www.ecmwf.int/en/forecasts/datasets/reanalysis-datasets/era5>. The near real
227 time MODIS land cover product was used to replace the outdated land cover in WRF
228 preprocessing system. We ran the model from before the onset of each snowfall event
229 until and after all snowmelt following the event had ceased. EXP1 was run Nov. 1-8
230 2018, EXP2 was run Nov. 13-18 2018, EXP3 was run Dec. 16-20 2018, EXP4 was run
231 Feb. 5-11 2019, EXP5 was run Mar. 2-10 2019, EXP6 was run Mar. 12-18 2019, EXP7
232 was run Mar. 24-29 2019, EXP8 was run Mar. 28-Apr.7 2019. The model results were
233 output at 3 hours intervals and the first day was used for model spin up.

234 **2.3 Data for model evaluation and comparison**

235 CMA hourly observations of air temperature and snow depth from 502 stations were

236 used to assess the WRF model estimates of 2 m air temperature and snow depth that
237 were made using the improved snow albedo parameterization scheme. Albedo is a key
238 factor for net radiation calculations, and is defined as the ratio of reflected shortwave
239 radiation (upwards) to received shortwave radiation (downwards). It determines the
240 distribution of turbulent land surface heat fluxes between sensible (SH) and latent heat
241 (LH). There are many meteorological observations available that have been
242 continuously recorded on the Tibetan Plateau at atmospheric boundary layer towers,
243 eddy covariance systems (Ma et al., 2018, 2020) and the Qilian Mountains integrated
244 observatory network (Li, 2019; Liu et al., 2020; Zhao and Zhang, 2020). These provide
245 in situ data that are assumed to constitute ‘truth’ for the model validation in this study.
246 In situ observations of albedo, SH and LH from 11 Chinese Academy of Sciences (CAS)
247 stations/samples, and from 16 stations in the Qilian Mountains integrated observatory
248 network are used to evaluate the accuracy of the improved snow albedo
249 parameterization scheme for modeling snow events on the Tibetan Plateau. It is
250 reasonable to compare observations of albedo, SH and LH with model estimates at 5
251 km resolution because there are only a few in situ observation stations in d02, but a
252 total of 27 observation stations in d01. At local solar noon in Lhasa (14:00 BST), the
253 observed albedo value is closer to the Lambertian albedo that is described by the WRF
254 model when coupled with LSMs. We therefore used albedo observations made at 14:00
255 BST to evaluate the model calculated albedo. Quality control codes 1, 2 and 3 were
256 selected when using the Turbulence Knight version 3 (TK3) software, and 0 was used
257 when using the Eddypro software to calculate SH and LH. Details of the 27 stations
258 from the CAS and the Qilian Mountains integrated observatory network that were used
259 in our study are provided in Table 2, and their locations on the Tibetan Plateau are
260 shown in Figure 1. *In order to compare WRF simulations against in situ observations,*
261 *sampling the gridded model estimates and interpolating to given ground stations’*
262 *locations were done by bi-linear interpolation of the four surrounding model grid points.*
263 The RMSE and the correlation coefficient were calculated for the assessment of the
264 model performance in relation to albedo, air temperature, SH, LH, and snow depth

265 estimates.

266 **Table 2** Location of stations from the Chinese Academy of Sciences (CAS) and the
267 Qilian Mountains integrated observatory network, and whether or not observations of
268 albedo, SH and LH were used from each station.

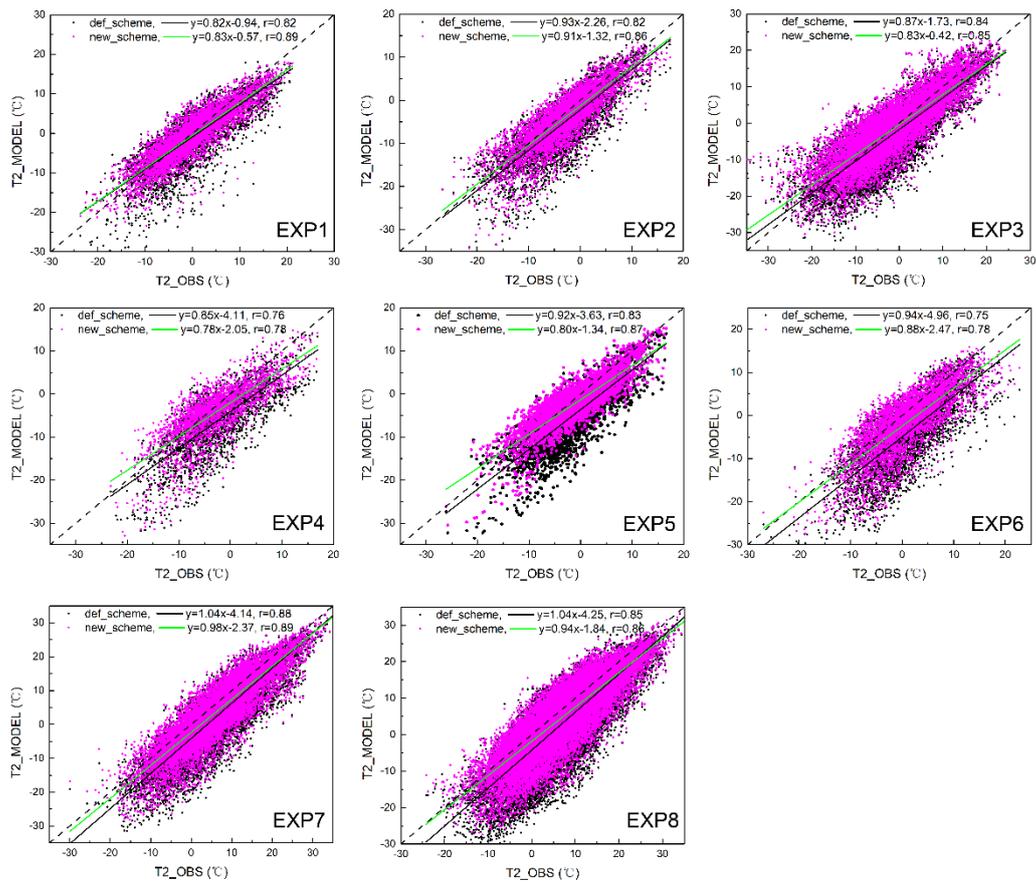
Station No.	Station Name	Station Type	Latitude (° N)	Longitude (° E)	Elevation (m)	Using albedo	Using SH and LH
1	SETS	CAS	29.77	94.73	3326	Yes	No
2	QOMS	CAS	28.21	86.56	4276	Yes	Yes
3	QOMS sample	CAS	28.31	86.85	4600	No	Yes
4	MASWE	CAS	38.41	75.05	3668	Yes	No
5	Nam Co	CAS	30.77	90.99	4730	Yes	Yes
6	NASDE	CAS	33.39	79.70	4264	Yes	No
7	Shuanghu	CAS	33.22	88.83	4947	Yes	Yes
8	NewD66	CAS	35.43	93.59	4465	Yes	No
9	BJ	CAS	31.37	91.90	4509	Yes	Yes
10	MS3478	CAS	31.93	91.71	4620	Yes	No
11	Amdo	CAS	32.24	91.62	4695	Yes	No
12	Yakou	Qilian	38.01	100.24	4148	Yes	Yes
13	Arou	Qilian	38.05	100.46	3033	Yes	Yes
14	Jingyangling	Qilian	37.84	101.12	3750	Yes	Yes
15	Dashalong	Qilian	38.84	98.94	3739	Yes	Yes
16	Heihe Remote Sensing	Qilian	38.83	100.48	1560	Yes	No
17	Huazhaizi Desert Steppe	Qilian	38.77	100.32	1731	Yes	Yes
18	Daman	Qilian	38.86	100.37	1556	Yes	Yes
19	Zhangye wetland	Qilian	38.96	100.45	1460	Yes	Yes
20	Guazhou	Qilian	41.41	95.67	2014	Yes	Yes
21	Dayekou	Qilian	38.56	100.29	2703	Yes	No
22	Dunhuang	Qilian	40.35	93.71	993	Yes	No
23	Liancheng	Qilian	36.69	102.74	2903	Yes	No
24	Linze	Qilian	39.24	100.06	1402	Yes	No
25	Sidalong	Qilian	38.43	99.93	3146	Yes	No
26	Xiyinghe	Qilian	37.56	101.86	3616	Yes	No

27	Tianjun	Qilian	37.70	98.61	3718	Yes	Yes
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269 **3 Results**

270 **3.1 Air temperature**

271 Albedo is a key factor in determining the net radiation received at the surface, which
272 determines the land surface energy balance and influences air temperature. Scatterplots
273 of the air temperatures estimated by the WRF model and observed at the CMA stations
274 are shown in Figure 2. *In all eight modeling experiments, implementing the improved*
275 *snow albedo scheme in the WRF model greatly reduces the cold bias that occurs when*
276 *the default Noah snow albedo scheme is used. Where the default Noah scheme results*
277 *in a warm bias at the observed lower air temperature for EXP1 and EXP3, however,*
278 *the improved albedo scheme does not improve the accuracy of the WRF estimates.*
279 *Compared with cold bias caused by the default Noah albedo scheme at the observed*
280 *lower air temperature, the improved snow albedo scheme results in a warm bias for*
281 *EXP2, EXP4 and EXP5.* On the whole, scatterplots comparing air temperature
282 observations from CMA station with WRF estimates made using the improved snow
283 albedo scheme estimates, show the data to be concentrated near the ideal fitting line,
284 where the model has exactly reproduced the observations. *Using the improved snow*
285 *albedo scheme results in a marked reduction in the cold bias for the WRF model*
286 *estimates for EXP1, EXP2, EXP4, EXP5 and EXP6, and the greatest reduction in the*
287 *cold bias, for all eight experiments, occurs for EXP6 (Fig. 2).*



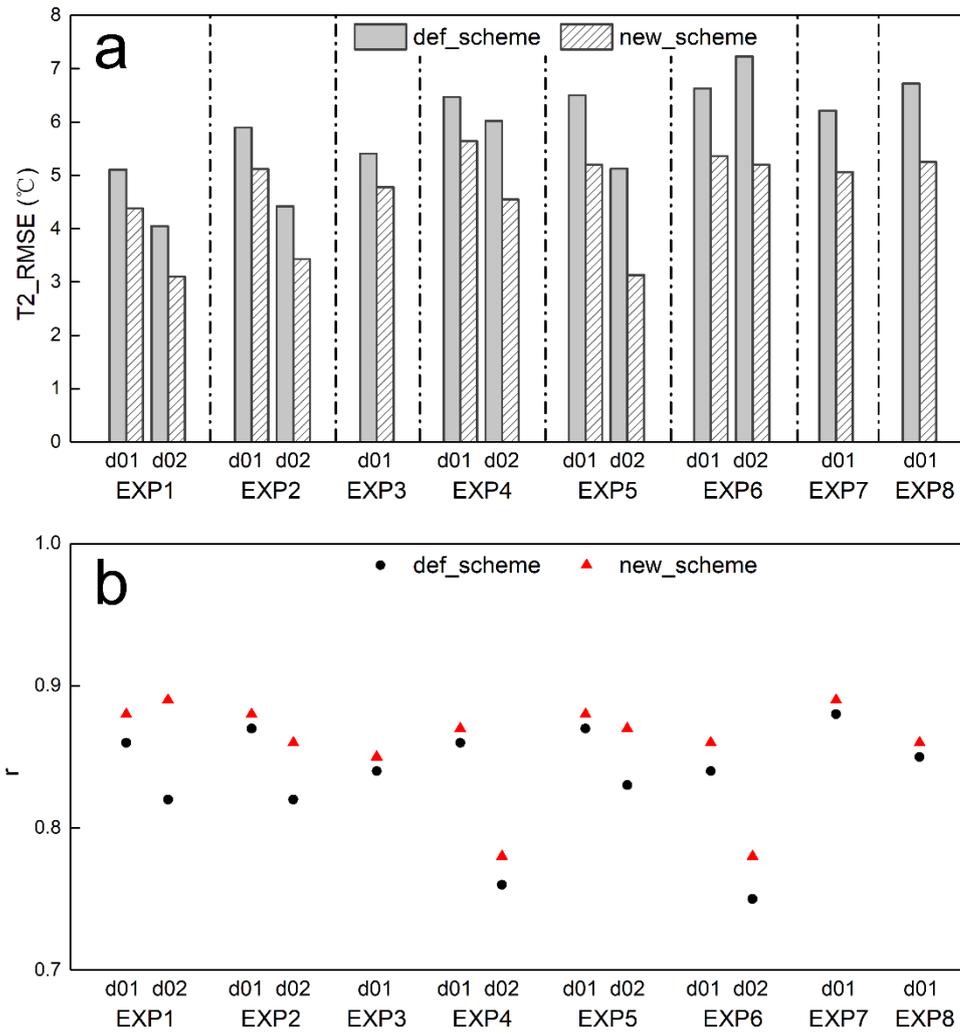
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289 **Figure 2.** Scatterplot of air temperature, comparing the CMA observations and model
 290 estimates for the eight experiments in the inner (high resolution) model domain from
 291 WRF, using the default Noah snow albedo scheme (def_scheme, black solid circle), and
 292 using the improved snow albedo scheme (new_scheme, red solid circle). The black
 293 solid line is a linear fit to the black solid circles. The green solid line is a linear fit to
 294 the red solid circles. The black dotted line is the line $y=x$. r is the correlation coefficient
 295 between the CMA observations and the model estimates. The correlation coefficient (r)
 296 is significant at 0.01 significance level.

297 To quantify the improvement to air temperature estimates that follows from
 298 implementation of the improved snow albedo scheme, the RMSE and correlation
 299 coefficient was calculated between CMA observations and model estimates of air
 300 temperature, shown in Figure 3. The differences between the accuracy of the new
 301 scheme and the default scheme are shown in Table 3. The accuracy of WRF air

302 temperature estimates varies between the different snowfall events, between the
303 different snow albedo schemes, and also varies with model resolution. The lowest air
304 temperature RMSE and the highest correlation coefficient are 3.1 °C and 0.89,
305 respectively, and both occur for EXP1. The highest air temperature RMSE and the
306 lowest correlation coefficient occur for EXP6, reaching 7.2 °C and 0.75, respectively.
307 Air temperature RMSE generally ranges from 4.1 to 7.2 °C for the WRF estimates that
308 were made using the default Noah snow albedo scheme estimates, with correlation
309 coefficients ranging from 0.75 to 0.88. In contrast, when the improved snow albedo
310 scheme is implemented in WRF, the RMSE ranges from 3.1 to 5.6 °C, and the
311 correlation coefficients range from 0.78 to 0.89. *Compared with when the default Noah*
312 *snow albedo scheme is used, the maximum decrease in air temperature RMSE when the*
313 *new scheme is used reaches 2.03 °C, which represents an improvement of 28.1 %*, and
314 the average decrease in air temperature RMSE is 1.2 °C, representing an improvement
315 of 20.7 %. There is an improvement of more than 11 % in the RMSE for all eight
316 experiments with the maximum improvement of 39 % when the new albedo scheme is
317 used, relative to when the default scheme is used. Implementing the improved snow
318 albedo scheme in WRF for all eight experiments also increased the correlation
319 coefficient between observed and modeled air temperature, by 0.01-0.07, which
320 represents an improvement of 1-9 % (Fig. 3, Table 3).

321 Compared with using the default Noah snow albedo scheme, using the improved
322 scheme results in improved model estimates for all eight EXPs, decreasing the air
323 temperature RMSE and increasing the correlation coefficient when compared with
324 observations. These improvements occur for air temperature estimates calculated at
325 both 5 km and 1 km resolution. The improvement to WRF model estimates is greater
326 for calculations made at 1 km resolution than at 5 km resolution, and air temperature
327 estimates are more accurate at 1 km resolution than at 5 km resolution *by implementing*
328 *the improved albedo scheme* (Fig. 3, Table 3). Therefore, fine resolution (i.e., 1 km) is
329 strongly recommended for future snowfall event modeling studies.



330

331 **Figure 3.** RMSE (a) and correlation coefficient (b) for air temperature (T2) between
 332 observations and model estimates in d01 and d02. The correlation coefficient is
 333 significant at the 0.01 significance level. For def_scheme and new_scheme, see Figure
 334 2.

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339 **Table 3** RMSE and correlation coefficient (r) for air temperature between CMA
340 observations and model estimates, calculated using the default Noah snow albedo
341 scheme (def_scheme) and the improved albedo scheme (new_scheme). The difference
342 in RMSE is new_scheme RMSE minus def_scheme RMSE. The difference in r is
343 new_scheme r minus def_scheme r. P value <0.05 is the significance test for the
344 correlation.

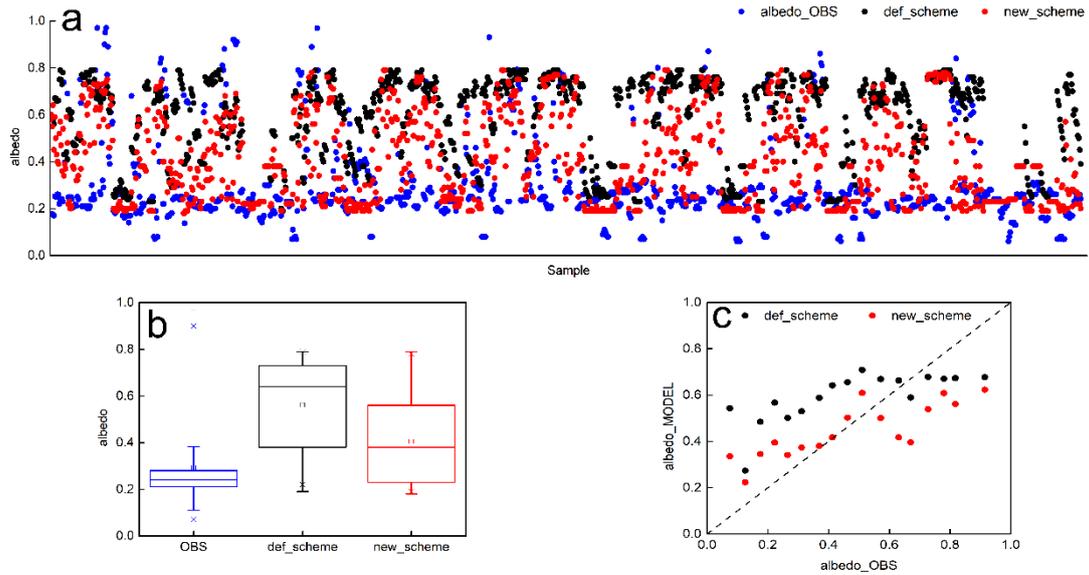
EXPs	Model Domain	RMSE (°C)				r			
		def_ scheme	new_ scheme	Difference	Relative Difference	def_ scheme	new_ scheme	Difference	Relative Difference
EXP1	d01	5.11	4.38	-0.73	-14.3 %	0.86	0.88	0.02	2.3 %
	d02	4.05	3.1	-0.95	-23.5 %	0.82	0.89	0.07	8.5 %
EXP2	d01	5.9	5.12	-0.78	-13.2 %	0.87	0.88	0.01	1.1 %
	d02	4.42	3.43	-0.99	-22.4 %	0.82	0.86	0.04	4.9 %
EXP3	d01	5.41	4.78	-0.63	-11.6 %	0.84	0.85	0.01	1.2 %
EXP4	d01	6.47	5.64	-0.83	-12.8 %	0.86	0.87	0.01	1.2 %
	d02	6.02	4.55	-1.47	-24.4 %	0.76	0.78	0.02	2.6 %
EXP5	d01	6.5	5.2	-1.30	-20.0 %	0.87	0.88	0.01	1.1 %
	d02	5.13	3.13	-2.00	-39.0 %	0.83	0.87	0.04	4.8 %
EXP6	d01	6.63	5.36	-1.27	-19.2 %	0.84	0.86	0.02	2.4 %
	d02	7.23	5.2	-2.03	-28.1 %	0.75	0.78	0.03	4.0 %
EXP7	d01	6.21	5.06	-1.15	-18.5 %	0.88	0.89	0.01	1.1 %
EXP8	d01	6.72	5.25	-1.47	-21.9 %	0.85	0.86	0.01	1.2 %

345 **3.2 Albedo**

346 Using the improved albedo scheme in the WRF model greatly reduces the cold air
347 temperature bias that otherwise occurs, indicating an improvement to model
348 performance. It is necessary to compare albedo estimates with in situ observations.
349 There are very few observation stations located in the finer model domain, and so a
350 total of 26 stations in d01 were used to evaluate the performance of the WRF

351 simulations of albedo at 5 km resolution (Table 2).

352 Scatterplots comparing observations and WRF estimates for albedo in the eight
353 experiments, when both the default and improved snow albedo scheme were used, are
354 shown alongside our statistical analysis in Figure 4. Albedo higher than 0.7 is
355 interpreted as snowfall. Albedo in the range of 0.4 to 0.6 is interpreted as snowmelt.
356 Albedo lower than 0.3 indicates sparse or patchy snow cover at the in situ stations. For
357 all eight snowfall events, the observed albedo is concentrated at low values, with a
358 median of 0.24, while WRF estimated albedo using the default Noah snow albedo
359 scheme has higher values, with a median of 0.64. Compared with albedo estimates
360 calculated using the default scheme, WRF estimates made using the improved scheme
361 result in a prolonged snowmelt period, which increases the number of snowmelt
362 samples and leads to a median albedo of 0.38, which is closer to that for the in situ
363 observations. The mean albedo estimated from WRF using the improved scheme is 0.4,
364 which is also closer to the observed mean of 0.3, than the mean of 0.6 calculated from
365 WRF using the default scheme (Fig. 4a, 4b). In general, the accuracy of the WRF
366 estimates when the new scheme is used is closely related to the observed albedo.
367 Compared with the WRF estimates made using the default Noah scheme, the WRF
368 estimates made using the improved scheme greatly reduce the overestimation of albedo
369 when the observed values are below 0.6, but seem to increase the underestimation of
370 albedo when the observed values are higher than 0.6 (Fig. 4c).

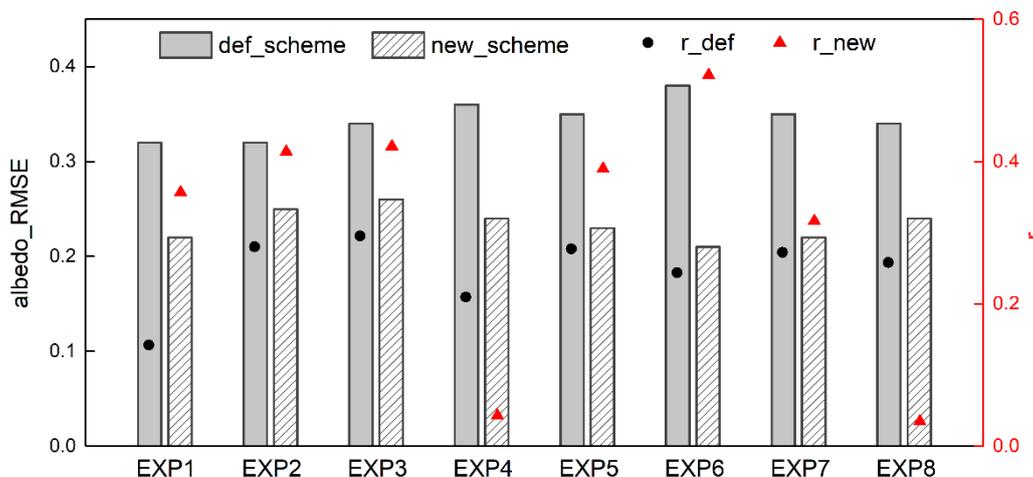


371

372 **Figure 4.** Scatterplot of albedo from observations and model estimates for the eight
 373 experiments (a: horizontal axis denotes samples from EXP1 to EXP8), corresponding
 374 box-and-whisker plot (b) *with lower and upper boundaries of the box indicating the*
 375 *first and third quartiles of albedo respectively and the line inside the box indicating the*
 376 *median of albedo, and averaged albedo observations every 0.05 segments (i.e., 0-0.049,*
 377 *0.05-0.099, 0.1-0.149, 0.15-0.199, ..., 0.85-0.899, 0.9-0.949, 0.95-0.999) and model*
 378 *estimates at the same time as observations* for all experiments (c). For def_scheme and
 379 new_scheme, see Figure 2.

380 To further evaluate the accuracy of WRF albedo estimates when the different snow
 381 albedo schemes are used, we calculated the RMSE and correlation coefficients between
 382 the observations and the model estimates (Figure 5). The RMSE for the WRF estimates,
 383 when compared to the observations, ranges from 0.32 to 0.38 for the eight experiments
 384 when the default scheme is used, and ranges from 0.21 to 0.26 when the improved
 385 scheme is used. Compared with the default Noah snow albedo scheme, the improved
 386 scheme results in a 0.1 decrease to the albedo RMSE in EXP1, representing a relative
 387 decrease of 31.3 %; a 0.07 decrease in EXP2, representing a relative decrease of 21.9 %;
 388 a 0.08 decrease in EXP3, representing a relative 23.5 % decrease; a 0.12 decrease in
 389 EXP4 and EXP5, representing relative decreases of 33.3 % and 34.3 %, respectively; a

390 0.17 decrease in EXP6, representing a relative 44.7 % decrease; a 0.13 decrease in
 391 EXP7, representing a relative 37.1 % decrease; and a 0.1 decrease in EXP8,
 392 representing a relative 29.4 % decrease. With the exceptions of EXP4 and EXP8,
 393 correlations between the modeled and observed albedo are significant at the 0.01
 394 significance level. Implementing the improved albedo scheme in WRF increases the
 395 albedo correlation coefficient by 0.21 in EXP1, a relative increase of 151 %; by 0.13 in
 396 EXP2, a relative increase of 47.6 %; by 0.13 in EXP3, a relative increase of 42.5 %; by
 397 0.11 in EXP5, a relative increase of 40.7 %; by 0.28 in EXP6, a relative increase of
 398 114 %; and by 0.04 in EXP7, a relative increase of 16.2 % (Fig. 5). In general, during
 399 snowfall and the snowmelt period that follows it, implementing WRF using the
 400 improved snow albedo scheme outperforms implementing WRF using the default Noah
 401 scheme and results in more accurate albedo estimates, demonstrated by considerable
 402 decreases in RMSE and increases in the correlation coefficients.

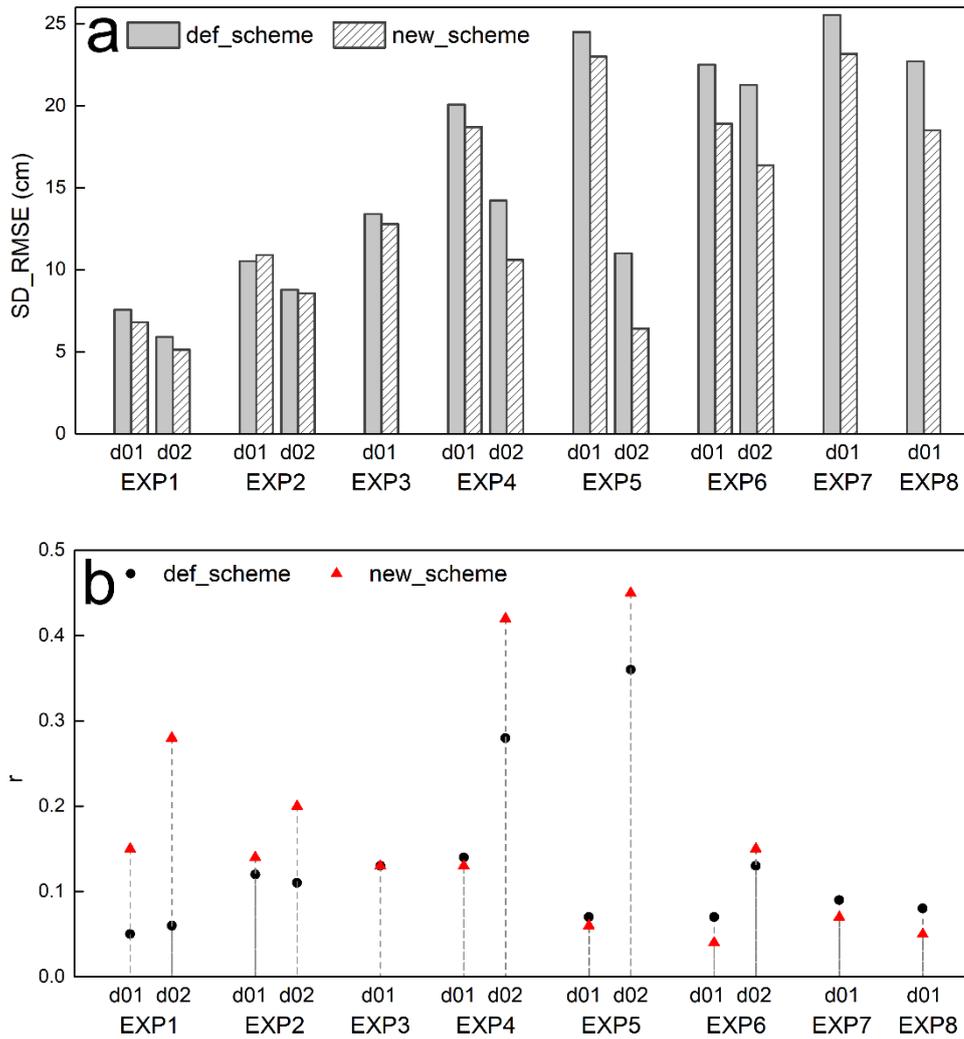


403

404 **Figure 5.** RMSE in column and correlation coefficient (r) in scatterplot for albedo,
 405 comparing observations and model estimates. r_def and r_new mean r between the
 406 observations and WRF-estimates using the default albedo scheme, and using the
 407 improved albedo scheme, respectively. Except EXP4 and EXP8, the correlation
 408 coefficient is significant at the 0.01 significance level. For def_scheme and
 409 new_scheme, see Figure 2.

410 **3.3 Snow depth**

411 There is a feedback between albedo and snow. Snow accumulation and snowmelt
412 influence the proportion of solar irradiance that is reflected; albedo indirectly and non-
413 negligibly influences snow accumulation and snowmelt by affecting the land surface
414 energy budget. In this study, we use an improved snow albedo scheme in which albedo
415 is parameterized as a function of snow depth and age. WRF model estimates of albedo
416 calculated using the improved snow albedo scheme outperform those calculated using
417 the default Noah scheme when snowfall events are simulated, and this leads to
418 improved representation of snowfall and the subsequent snowmelt processes in WRF
419 when the improved scheme is used. Instantaneous direct measurements of snow depth
420 are recorded during snowfall events and over the subsequent snowmelt period. We use
421 these to quantify the improvement that using the new albedo scheme makes to snow
422 estimates calculated in WRF. We assess this by calculating the RMSE and correlation
423 coefficient between the model snow depth estimates and CMA observations, as shown
424 in Figure 6.



425

426 **Figure 6.** Same as Figure 3, but for RMSE (a) and correlation coefficient (b) for snow
 427 depth (SD). The correlation coefficient is significant at the 0.01 significance level,
 428 except for d01 estimates in EXP6.

429 Comparing the accuracy of WRF snow depth estimates calculated using the new
 430 scheme, to the accuracy achieved using the default Noah scheme, the greatest relative
 431 decrease in RMSE is 41.7 %, which occurs for estimates made at finer resolution in
 432 EXP5. Replacing the default albedo scheme with the new scheme in WRF results in an
 433 average decrease in snow depth RMSE of 2.2 cm, which is a 13.4 % improvement. In
 434 areas covered by the higher resolution model domain, the average RMSE decrease is
 435 2.8 cm, which is a 21.2 % improvement. This shows that the impact of replacing the

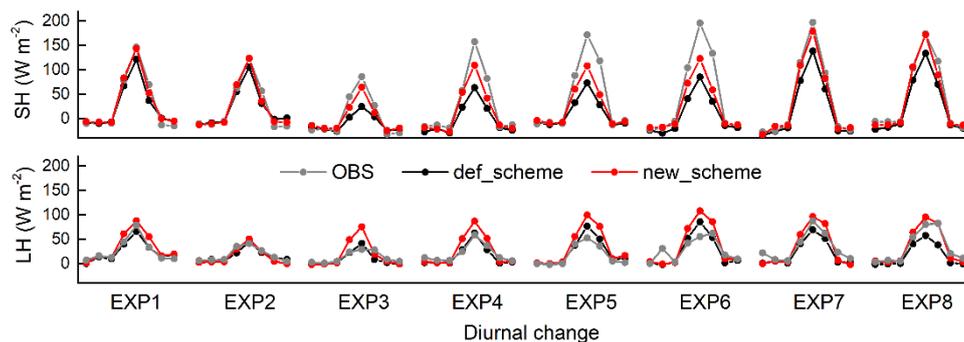
436 albedo scheme with an improved scheme is more significant for areas in the higher
437 resolution d02 model domain than for areas in the coarser d01 model domain (Fig. 6a).
438 Using the improved albedo scheme in WRF increases the correlation coefficient
439 between observed and modeled snow depth in areas within the d02 model domain, but
440 this increase is not consistent for areas in the d01 domain. The greatest increase, both
441 relative and absolute, in the snow depth correlation coefficient occurs in the finer
442 simulation domain in EXP1, where the correlation between observed and modeled
443 snow depth increases by 0.22, which is a 366.7 % increase. The mean and relative
444 increases in the correlation coefficient between observed and modeled snow depth for
445 areas in the d02 simulations are 0.14 and 107.8 %, respectively (Fig. 6b). WRF snow
446 depth estimates are more accurate at finer resolution (i.e., 1 km resolution) than in
447 coarser simulations (i.e., 5 km resolution), regardless of which albedo scheme is
448 implemented (Fig. 6). Implementing the improved albedo scheme in WRF improves
449 the agreement between model estimated and observed snow depth, relative to
450 implementing the default Noah albedo scheme, as seen by the decreased RMSE and the
451 increased correlation coefficient.

452 **3.4 Turbulent heat and vapor fluxes**

453 Albedo plays a significant role in the land surface energy balance. It determines the
454 proportioning of net radiation fluxes between turbulent heat and vapor fluxes. Our study
455 shows that using the improved snow albedo scheme in WRF results in a good model
456 representation of surface albedo in simulations of snow events. We now consider
457 whether replacing the default scheme with the improved scheme may affect the
458 proportioning of net radiation fluxes between turbulent heat and vapor fluxes. Since
459 there are very few observation in situ stations located in the area covered by the finer
460 model domain, a total of 14 in situ stations located in the area covered by d01 were
461 selected, and WRF estimates of turbulent heat and vapor fluxes were assessed from
462 simulations calculated at 5 km resolution (Table 2)

463 The diurnal changes in SH and LH recorded in the in situ observations and calculated

464 in the eight modeling experiments are shown in Figure 7. The WRF model successfully
 465 captures the diurnal changes in SH and LH, particularly in EXP1 and EXP2, where the
 466 model estimates of SH and LH are almost equal to the in situ observations. In the
 467 nighttime, the WRF model accurately estimates SH and LH in all eight experiments.
 468 However, during the day WRF consistently underestimates SH in all experiments
 469 except EXP1 and EXP2, and estimates LH with varying accuracy when the default
 470 Noah albedo scheme is used. For example, when the default Noah scheme is used, WRF
 471 accurately estimates LH in EXP3 and EXP4, but overestimates LH in EXP5 and EXP6
 472 and underestimates LH in EXP7 and EXP8. Compared with WRF estimates calculated
 473 using the default scheme, WRF simulations calculated using the new albedo scheme
 474 result in increased estimates of the turbulent heat and vapor fluxes. This leads to SH
 475 estimates that are closer to observations for experiments EXP3 to EXP8, a greatly
 476 overestimated LH for experiments EXP3 to EXP6, and a slightly overestimated LH,
 477 which is closer to observations for EXP7 and EXP8 (Fig. 7). In general, WRF estimates
 478 of SH are improved through the implementation of the new albedo scheme, relative to
 479 the default scheme, although SH is underestimated during snowfall events. The impact
 480 of the improved albedo scheme on LH estimates varies between the different snowfall
 481 events, but LH is consistently overestimated.

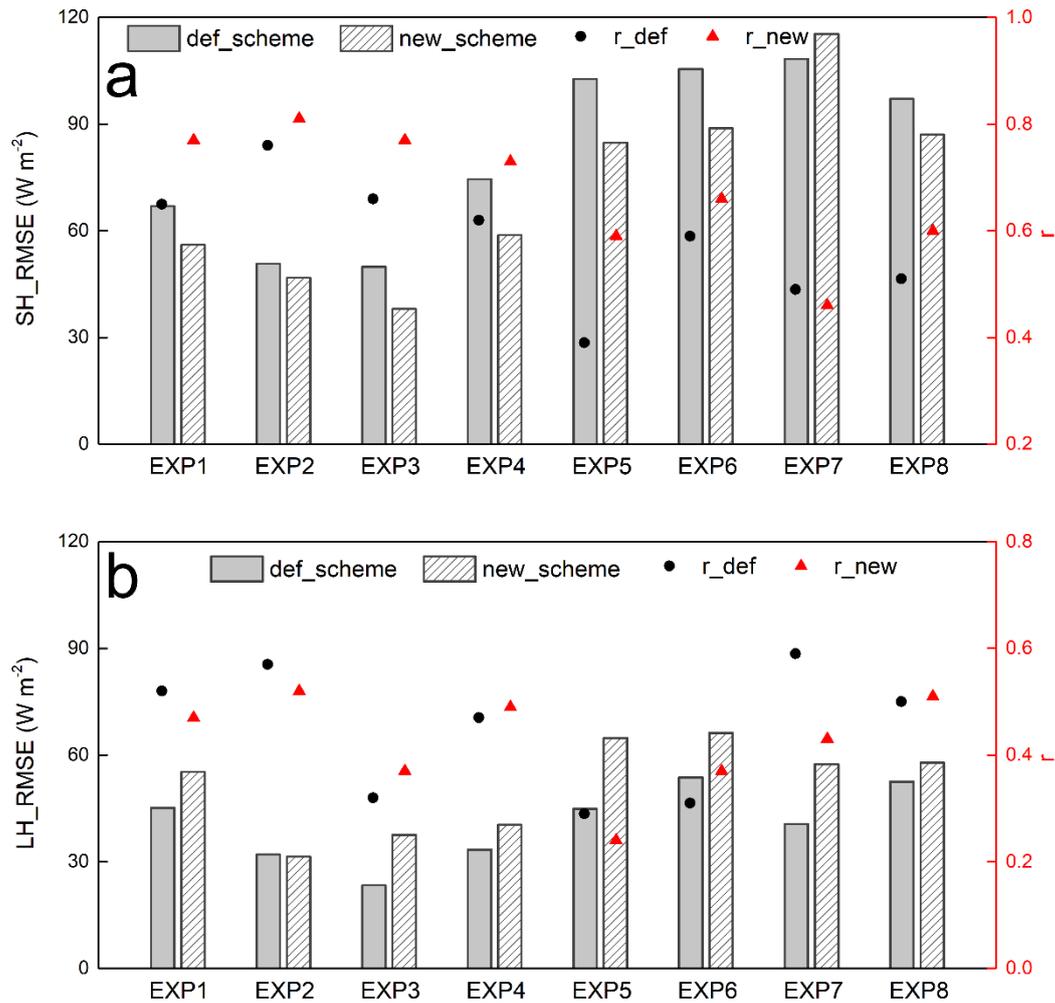


482

483 **Figure 7.** Diurnal change of sensible (SH) and latent (LH) heat fluxes from ground
 484 observations and model estimates. Beginning at 02:00 BST with 3 hours interval. For
 485 def_scheme and new_scheme, see Figure 2.

486 The RMSE and correlation coefficient for comparisons between WRF estimates and
487 observations for SH and LH are shown in Figure 8. Compared with when the default
488 Noah scheme is used, WRF estimates using the improved scheme result in reduced
489 RMSE for SH estimates in all experiments except EXP7. The absolute (and relative)
490 reductions are: 10.9 W m⁻² (16.2 %) in EXP1, 3.9 W m⁻² (7.6 %) in EXP2, 11.9 W m⁻²
491 (23.8 %) in EXP3, 15.7 W m⁻² (21 %) in EXP4, 17.9 W m⁻² (17.4 %) in EXP5, 16.6 W
492 m⁻² (15.8 %) in EXP6, and 10.0 W m⁻² (10.3 %) in EXP8. There is a relative increase
493 of 6.5 % in EXP7. Implementing the improved scheme in WRF significantly increases
494 the correlation coefficients between observed and modeled SH, relative to using the
495 default scheme, in all experiments except EXP7, where there is a slight decrease. The
496 largest increase in the SH correlation coefficient, both absolute and relative, is 0.2 and
497 51.3 %, respectively, and occurs for EXP5. Implementing the improved scheme in WRF
498 reduces the SH RMSE by an average of 10 W m⁻², which is an improvement of 13.2 %
499 improvement, and increases the SH correlation coefficient by an average of 0.1, which
500 is an improvement of 16.8 % (Fig. 8a). However, replacing the default scheme with the
501 improved albedo scheme results in less accurate estimates of LH, and corresponds to
502 an increase in RMSE in almost all eight experiments, although the correlation
503 coefficient increases for half of the snowfall events simulations (EXP3, EXP4, EXP6
504 and EXP8) (Fig. 8b).

505 In summary, the improved snow albedo scheme has a significant effect on the
506 proportioning of radiative fluxes between turbulent heat and vapor fluxes. It
507 significantly outperforms the default Noah scheme in relation to SH estimates, but there
508 is no significant improvement in LH estimates and these may be less accurate when the
509 new scheme is used, relative to the default scheme, during snowfall and the subsequent
510 snowmelt period.



511

512 **Figure 8.** Same as Figure 5, but for sensible (a) and latent (b) heat fluxes. The
 513 correlation coefficient is significant at the 0.001 significance level.

514 **4 Discussion**

515 The highly complex topography of the Tibetan Plateau means that WRF estimates of
 516 air temperature, albedo and snow depth are strongly sensitivity to model resolution. Our
 517 study shows that WRF performs much more accurately when run at finer resolution (1
 518 km) than at relatively coarse resolution (5 km) for snowfall events over the Tibetan
 519 Plateau, regardless of which snow albedo parameterization scheme is used. This
 520 difference may be explained by the ability of the model to resolve complex terrain
 521 (Rahimi et al., 2019), and/or by the implementation of the cumulus convective

522 parameterization scheme. The more detailed representation of complex terrain and the
523 explicit representation of convection mean that running WRF at finer resolution greatly
524 improves model estimates of air temperature, surface pressure, and relative humidity
525 (Singh et al., 2020), and provides small improvements in the magnitude of daytime
526 convective precipitation (Collier and Immerzeel, 2015). Norris et al. (2017) pointed out
527 that decreasing the grid spacing from 6.7 to 2.2 km likely improves estimates of
528 mountain precipitation but does not fundamentally change the representation of the
529 diurnal cycle. They indicated that the key difference between low and high model
530 resolution is whether or not a cumulus convective scheme is required. Subkilometer
531 grid resolution has been investigated in WRF, and used for modeling meteorological
532 variables over complex terrain (Horvath et al., 2012; Dimitrova et al., 2016). The 500
533 m resolution configuration of WRF results in the closest match between the model
534 estimates and observations, and gives the most plausible spatial distribution of
535 precipitation over the complex topography. The performance of the WRF model has
536 been similarly demonstrated to improve at 500 m, relative to coarser resolutions, for
537 wind and air temperature estimates (Bonekamp et al., 2018). We therefore strongly
538 suggest that subkilometer grid resolution should be considered when WRF is
539 configured for simulations covering areas in High Mountain Asia in future research.

540 Our improved snow albedo scheme parameterizes albedo as a function of snow depth
541 and age by considering the relationship between MODIS albedo and the modelled snow
542 depth and age. It is more physically plausible than the default Noah scheme, which
543 considers snow cover, and outperforms the default Noah scheme for air temperature,
544 snow depth, albedo and turbulent heat and vapor fluxes estimates during snowfall
545 events. However, even when the improved albedo scheme is used, the RMSE for WRF
546 estimates of albedo at 5 km spatial resolution remains around 0.21-0.26, although this
547 represents a decrease of 22-45 % relative to when the default albedo scheme is used. It
548 should be noted that the accuracy of the MODIS albedo retrieval algorithm is limited
549 during snowfall events and snowmelt periods (Qin et al., 2011; An et al., 2020), and
550 also that rugged mountain terrain not only affects the radiation absorbed by the land

551 surface, but also affects the radiation reflected by the land surface to the satellite borne
552 sensor. Multiple reflection and scattering from adjacent mountains creates challenges
553 for the monitoring and retrieval of surface albedo in areas of complex terrain via remote
554 sensing (Zhang and Gao, 2011; Roupioz et al., 2014, 2016). This reduces the accuracy
555 of MODIS albedo retrieval over the complex topographic Tibetan Plateau and
556 constitutes a limitation to the improved albedo parameterization investigated here, since
557 it relies on MODIS albedo products. A terrain correction is required for the MODIS
558 albedo retrieval to further improve the albedo parameterization scheme used here.
559 However, it is difficult to establish a unified terrain correction model due to the large
560 undulations of the Tibetan Plateau. How to effectively eliminate the influence of terrain
561 factors from a specific mountain surface on quantitative retrievals from remote sensing
562 data has long been a challenge and a focus for remote sensing research. A further
563 challenge for the assessment presented here is the sparse and uneven distribution of
564 available in situ albedo observation data over the bulk Tibetan Plateau. This paucity of
565 data means that there is a mismatch in spatial resolution when comparing albedo
566 estimates, calculated at 5 km resolution, with in situ observations.

567 Air temperature is a critical factor that is related to albedo and determines energy
568 distribution between SH and LH. Using the improved snow albedo parameterization
569 scheme significantly reduces the albedo overestimates during snowfall events that
570 occur when the default scheme is used, and this leads to the reduction of the cold air
571 temperature bias in the model. Therefore, in this study, the improved scheme reduces
572 the underestimates for SH and improves the performance of WRF for simulating SH
573 over the Tibetan Plateau, relative to when the default scheme is used. These results
574 indicate that the accurate simulation of surface albedo is very important for the accurate
575 simulation of SH.

576 Implementing the improved albedo scheme results in little improvement to estimates of
577 LH calculated, which is restricted by water content and increased only slightly, relative
578 to when the default albedo scheme is used. This may be explained by the LH

579 parameterization scheme used in the Noah LSM (Chen and Dudhia, 2001). The total
580 LH in the Noah LSM has three components (LH from the direct evaporation from the
581 top surface layer, evaporation of precipitation intercepted by the canopy and
582 transpiration via the canopy and roots respectively). The factors affecting calculation
583 of LH in the Noah LSM include not only the radiation balance (which is impacted by
584 albedo), but also soil water, soil capillary conductivity and vegetation status, i.e., albedo,
585 surface heat and water vapor exchange coefficient, saturated water vapor pressure,
586 specific humidity, surface soil water content, field capacity, wilting point, canopy
587 resistance, total precipitation, and canopy interception amount. In our current study, we
588 have focused on the snow albedo parameterization scheme in the Noah LSM by
589 considering the MODIS albedo product and the additional snow related variable of
590 snow depth. Therefore, the influence of our improved scheme on LH estimates
591 calculated by the LSM is very limited.

592 Implementing the improved snow albedo scheme in place of the default scheme greatly
593 decreases the overestimation of albedo from snowmelt to snow free processes, but does
594 not remove the underestimation of albedo during snowfall. This means that the
595 improvements mainly come from snowmelt and snow free simulations, and model
596 performance during snowfall may be worse when the improved albedo scheme is used.
597 This suggests an opportunity to further investigate how albedo is characterized by snow
598 depth and age in the snow albedo parameterization scheme.

599 **5 Conclusions**

600 We conducted several numerical experiments to evaluate the performance of the
601 MODIS albedo based snow albedo parameterization scheme (Liu, 2020) implemented
602 in WRF. We assessed the RMSE and correlation coefficient between observed and
603 modeled air temperature, albedo, snow depth and turbulent heat and vapor fluxes for
604 simulations of eight snowfall events over the Tibetan Plateau. We compared the
605 accuracy of WRF estimates made using the improved snow albedo scheme with that of
606 WRF estimates made using the default Noah scheme, in both cases comparing with

607 ground observations.

608 The accuracy of WRF estimates of albedo is significantly improved when the new
609 albedo scheme is implemented. The default Noah scheme tends towards higher albedo
610 estimates and cannot accurately capture snowfall and snowmelt processes, resulting in
611 a high RMSE and low correlation coefficient between modeled and observed albedo.
612 Through consideration of snow related variables, such as snow depth and age, and by
613 being based on MODIS remote sensing albedo products, the improved scheme
614 estimates albedo more accurately than the default scheme, improving the albedo RMSE
615 by around 22 % to 45 %, with an average improvement of 32 %. Similarly, the improved
616 scheme results in an increased correlation coefficient between modeled and observed
617 albedo. Relative to the default scheme, the correlation coefficient relatively increases
618 by around 16 % to 151 %, with an average improvement of 69 %. This may contribute
619 to the relatively better performance of WRF for simulating air temperatures when the
620 improved albedo scheme is used. The improved scheme relatively decreases (increases)
621 the air temperature RMSE (correlation coefficient) by 16 % (1.5 %) for model estimates
622 calculated at 5 km resolution, and by 27 % (5 %) for model estimates calculated at 1
623 km resolution.

624 There are mutual feedbacks between snow and albedo. During snowfall and over the
625 subsequent snowmelt period, snow depth and age affect changes in the albedo. The
626 changes in albedo in turn affect the evolution of snow events by changing the surface
627 energy budget and the proportional distribution of net radiation between turbulent heat
628 and vapor fluxes, and finally by changing the type of precipitation. Our study shows
629 that when the default albedo scheme is replaced by the improved albedo scheme in
630 WRF, the turbulent heat and vapor fluxes estimates increase. The improved scheme
631 significantly outperforms the default Noah scheme for SH estimates, with a reduction
632 (increase) in the RMSE (correlation coefficient) of 10 W m^{-2} (0.1), representing an
633 improvement of 13 % (17 %). The overall accuracy with which WRF estimates
634 turbulent heat and vapor fluxes improves when the improved albedo scheme replaces

635 the default scheme, although there is no significant improvement in LH estimates. This
636 overall improvement leads to a more accurate reproduction of the evolution of snowfall
637 events and to more accurate snow depth estimates. Our study shows that using the
638 improved albedo scheme in WRF reduces the RMSE and increases the correlation
639 coefficient between modeled and observed snow depth, relative to using the default
640 scheme. This improvement is more significant for simulations at 1 km resolution than
641 for simulations at 5 km resolution, with maximum and averaged relative RMSE
642 (correlation coefficient) decreases (increases) of 42 % (367 %) and 21 % (108 %),
643 respectively for 1 km resolution simulations.

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660 observations from CAS and Qilian Mountains integrated observatory network is
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