

Analysis of atmospheric thermodynamics using the R package *aiRthermo*

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Abstract

The publicly available R package *aiRthermo* is presented in this study, which allows the user to process information relative to atmospheric thermodynamics, ranging from calculating the density of dry or moist air and converting data between moisture indices to processing a full sounding, obtaining factors such as the convective available potential energy, additional instability indices, or adiabatic evolutions of particles. The package also provides the possibility to present information using customisable Stüve diagrams. Many of the functions are written inside a C extension to ensure that the computations are fast. The results of applying this package to five years of real soundings measured over the Iberian Peninsula are also presented as an example. The package considerably extends the capabilities of R to process atmospheric soundings or model results. This will be useful for many practical environmental forecasting applications at different scales, such as statistical downscaling for climate analysis, quantitative precipitation forecasting (particularly precipitation extremes), diagnosing storms, flash floods, and lightning, and in aviation and other fields where computing atmospheric convection and its related parameters is important.

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Authorship statement: Jon Sáenz thought up the original concept of the package, wrote most parts of the C code, several parts of the R code, and some of the verification

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routines. He was also the lead author of the paper. Santos J. González-Rojí wrote some parts of the R and verification codes, packaged the software for its inclusion in CRAN, and wrote some parts of the paper. Sheila Carreno-Madinabeitia was the author of the Stüve plotting routines and parts of the verification code. She analysed the sounding database presented in the paper and participated in the writing. Gabriel-Ibarra Berastegi was involved in the verification of results, the testing of the package, and the writing of the paper.

Declarations of interest:

The only interest of the authors is to make the package known. If the paper is accepted, their scientific merit will also be publicly recognised.

Highlights

1. A new R-package providing new functions in atmospheric thermodynamics is presented.
2. The package provides analyses and Stve diagrams not previously available in R.
3. A case-study (2010-2014) of soundings over the Iberian Peninsula is presented.
4. The package can play a key role in advanced forecast and diagnostic systems.
5. The most critical functions are written in C to speed up computation.

1. Introduction

R (<https://www.cran.r-project.org/>) is a freely available software for statistical computing that has expanded exponentially in recent years. Some packages at the intersection between air pollution, climate, and atmospheric studies make R a powerful tool for processing data and visualising atmospheric processes, such as air pollution [1] and hydrology [2], and mapping [3] processing satellite [4] and atmospheric data [5]. Additional useful packages in this field, such as *RAtmosphere*, *ClimDown*, *opentraj*, and others, can be downloaded from the CRAN repository.

However, the analysis of atmospheric soundings and water vapour from the perspective of atmospheric thermodynamics is an area where the community of atmospheric scientists have new needs using R, as identified by the authors. *aiRthermo* extends the functionality offered by the *RadioSonde* and *meteogRam* packages by adding Stüve diagrams and the vertical evolution of air parcels.

Under the typical pressure and temperature conditions found in the atmosphere, the state of dry air is commonly described by two thermodynamic variables (pressure P and temperature T). The concentration of water can be expressed using different moisture indices, such as the specific humidity, mixing ratio, virtual temperature, and relative humidity. Phase changes of water cause large latent heat fluxes that must be considered as air parcels ascend or descend [6, 7, 8] when studying atmospheric stability and convection.

Some instability indices are often used when diagnosing meteorological situations. For example, the relationship between sea breeze and precipitation over Hainan Island [9], the role of moist convection in the development of flash floods [10], or the retrieval of precipitation in the Tropical Rainfall Measuring Mission (TRMM) mission [11], to name a few. Thermodynamic variables and indices are also used for the statistical downscaling of extreme precipitation events and moisture transport [12].

The main objective of this paper is to present an R package that has been designed to allow scientists to conduct computations involving atmospheric thermodynamics using the R language. This considerably extends the capabilities of R for meteorological data analysis, and has, for instance, allowed us to extend the lectures offered at the M.Sc. level [13], enabling students to perform

34 numerical exercises involving with these parameters.

35 **2. Data**

36 *2.1. Sample data provided with the package*

37 The data used in Section 4 of this paper (provided in the package for testing
38 by users) were collected from a server located at the University of Wyoming
39 (publicly available at <http://weather.uwyo.edu/upperair/sounding.html>).
40 The first sounding used in this paper (sounding A) was measured at Santander,
41 Spain (station ID 08023, date 2010-06-16, 12:00 UTC), and corresponds to a
42 day that faced some frontal rain. The second case (sounding D) was measured
43 at Barcelona, Spain (station ID 08190, 2013-08-07, 12:00 UTC) and reflects a
44 situation with substantial convective instability. The final example (Davenport,
45 USA, station ID 74455, 1997-06-21 at 00 UTC) is also used to illustrate a case
46 with strong convection [14].

47 *2.2. Data for the case study*

48 The topography of the Iberian Peninsula and the positions of the eight sound-
49 ing sites are shown in Figure 1. The sounding files downloaded from Wyoming
50 University’s server covered the period of 2010-2014. The number of cases de-
51 pends on the site, and ranges from 1705 over Lisbon to 3575 over Murcia. The
52 instability index values computed at Wyoming University and those computed
53 using the functions in *aiRthermo* were compared. In *aiRthermo*, the initial con-
54 ditions for CAPE were obtained by vertically averaging the lowest 500 m of
55 the sounding and by performing isobaric precooling for the resulting low-level
56 average particle values.

57 **3. Methodology**

58 Most of the theory and methods used to develop *aiRthermo* can be found
59 in standard references [6, 7, 8, 14, 15]. However, some of our assumptions are
60 documented below.

61 The state of an air parcel is defined in *aiRthermo* using its pressure P (Pa),
62 temperature T (K), and mixing ratio w (kg kg^{-1}). To compute the saturation

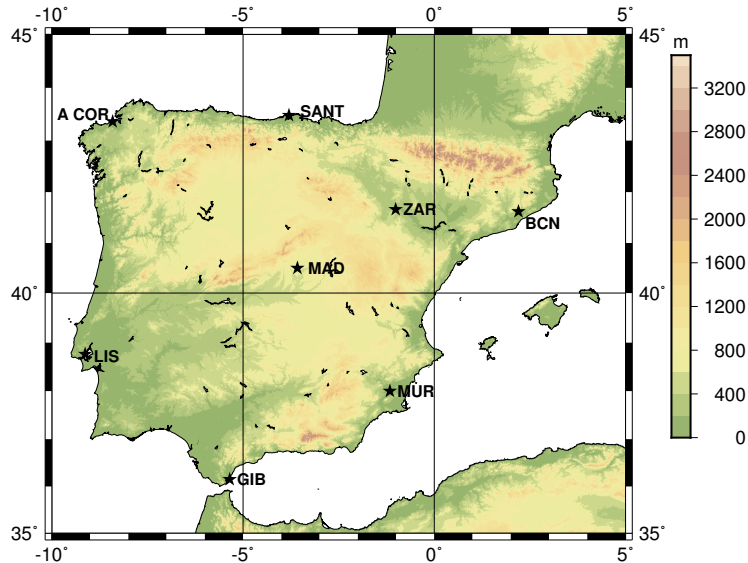


Figure 1: Map of the Iberian Peninsula showing the stations where soundings are routinely measured.

63 pressure of water over a flat surface, we follow the expressions on pages 197-200
 64 from [6] for ice and water below 30°C, and Buck’s equation [16] above 30°C.
 65 The dew-point temperature is given by the approximate expression of 5.68 in [6]
 66 from the given P and mixing ratio w .

67 The specific heat of moist air at a constant pressure and volume are com-
 68 puted as $c_{pm} = c_{pd}(1 + 0.87q)$ and $c_{vm} = c_{vd}(1 + 0.97q)$, following [14], with
 69 $c_{pd} = 1005 \text{ J kg}^{-1} \text{ K}^{-1}$ and $c_{vd} = 718 \text{ J kg}^{-1} \text{ K}^{-1}$ [6, 7]. By default, the functions
 70 that provide these specific heats will return the values corresponding to moist
 71 air, and values corresponding to dry air can be requested by explicitly building
 72 a dry air parcel.

73 A good analytical expression for the latent heat of evaporation of water
 74 does not exist, therefore, an approximate expression is used. For liquid water
 75 ($T \in [233.15, 313.15] \text{ K}$ interval), we use cubic polynomial expansion based on
 76 tabulated values [15] with an absolute residual smaller than 1 J kg^{-1} , and all
 77 terms are statistically significant at the 99% confidence level. By doing this, we
 78 assume that super-cooled water can exist up to the Schaefer point (about -40°C).
 79 The result for ice is based on a quadratic fit to the observed values [17] in the
 80 interval of $T \in [210, 273.15] \text{ K}$ (residuals smaller than 0.05 J kg^{-1}). As the values

81 corresponding to water and ice differ in the common interval used in the previous
 82 two expressions, a linear combination of ice's (L_i) and water's (L_w) latent heat
 83 is computed $L = w_i L_i + (1 - w_i) L_w$ with a weight of $w_i = 1 - \frac{T - 253.15}{20}$ in the
 84 interval of $T \in [253.15, 273.15]$ K. Below 253.15 K, the latent heat corresponding
 85 to ice is applied, and that corresponding to water is used above 273.15 K.

86 All vertical evolutions are computed assuming that a hydrostatic balance is
 87 in place. Therefore, $\Gamma_d^* = \left(\frac{dT}{dP}\right)_d = \frac{\Gamma_d}{\rho g}$ and $\Gamma_s^* = \frac{\Gamma_s}{\rho g}$ are used, with Γ_d and Γ_s
 88 as the typical expression for vertical gradients in Z . For the saturated pseudoa-
 89 diabatic profile, the expression used for the pressure coordinates is transformed
 90 from the common expression provided in z [6, 7, 8, 14]. Using the expressions
 91 of Γ_d^* (Γ_s^*) for dry (saturated) adiabatic evolutions, the vertical evolution of
 92 an air parcel from the initial state P_0 , T_0 , and w_0 is computed by numerically
 93 solving the ordinary differential equation $\frac{dT}{dP} = \Gamma_i^*(P, T, w)$, with $i = d$ or $i = s$
 94 depending whether or not the particle is saturated. This differential equation is
 95 numerically solved using the fourth-order Runge-Kutta scheme for all vertical
 96 evolutions, with saturation checks at each vertical step.

97 For some of the indices, such as the Lifted Index (LI) or Convective Available
 98 Potential Energy (CAPE), it is customary to calculate the vertical average of
 99 the lower levels to identify a representative parcel P_0 , T_0 , w_0 of the lowest levels
 100 of the atmosphere [18, 19]. These vertical averages at low levels are evaluated
 101 in all cases by first considering

$$\Delta Z = \frac{R_d}{g} \int_{P_t}^{P_s} \frac{T_v dP}{P} \quad (1)$$

102 as the vertical width of the parcel. Next, the accumulated vertical quantity
 103 value $X(P)$ is given by

$$\bar{X} = \frac{1}{\Delta Z} \frac{R_d}{g} \int_{P_t}^{P_s} \frac{X k T_v}{P} dP \quad (2)$$

104 where $k(P)$ serves as a normalising function. When calculating the average
 105 temperature, $k(P) = 1$ is used. In contrast, the specific humidity $k(P) = q(P)$
 106 is used to indicate moisture. The vertical integrals are computed using discrete
 107 slabs defined by the data given by the soundings in all cases. For these discrete
 108 slabs, the integrals are computed analytically and the results are accumulated.

109 To compute CAPE and convective inhibition (CIN), the vertical integrals
 110 are computed in pressure levels by adding the energy corresponding to discrete

111 slabs defined by linear or logarithmic vertical profiles, which are defined by the
112 soundings. The integrals for each of the slabs enclosed by linear profiles are com-
113 puted analytically, and the energy corresponding to each slab is accumulated,
114 producing the final value of CAPE or CIN. The integrals are always calculated
115 using the virtual temperature [20].

116 There are different methods of accurately determining the lifting conden-
117 sation level (LCL) or the equivalent potential temperature of an air parcel in
118 *aiRthermo*. In the first case, the package calculates these variables by comput-
119 ing their vertical evolutions and numerically solving the ordinary differential
120 equation representing their ascent from the initial conditions given by their
121 temperature, pressure, and mixing ratio. For compatibility, functions that al-
122 low these variables from well-known alternative equations to be computed, such
123 as the approximate method presented by Bolton [21] to compute LCL, are also
124 provided.

125 The routine designed to produce Stüve diagrams for either the soundings or
126 the lifted particles used in the computation of CAPE/CIN extends the available
127 options for thermodynamic diagrams in the existing RadioSonde R package. The
128 routine that plots the Stüve diagram uses the equivalent potential temperature
129 lines, constant mixing ratio lines, or dry adiabatic lines produced by the routines
130 in *aiRthermo* to ensure the full consistency of results. The routine that plots the
131 Stüve diagram allows the user to plot additional lines in the sounding, enabling
132 the production of highly customisable plots.

133 4. Description of the package

134 The package contains over 40 functions that can be separated into six large
135 groups according to their functionality. A brief description of each set of func-
136 tions is presented here, however, the manual of the package must be checked
137 for a full description of the functions and the parameters required to run them.
138 The manual can be found on the web-page for the package *aiRthermo* in the
139 Comprehensive R Archive Network (CRAN).

140 • Density of dry/moist air and virtual temperature

141 The density of air can be directly calculated using the corresponding func-
142 tion `densityMoistAir`, or by using intermediate functions to indepen-

143 dently calculate the densities of dry air and water vapour (`densityDry`
144 and `densityH20v`). The virtual temperature of an air parcel can also be
145 directly calculated by using the `virtual_temperature` function.

146 • Conversion of moisture indices

147 Several functions that allow conversion between moisture indices through
148 the dew point temperature, mixing ratio, and specific or relative humidity
149 are included in *aiRthermo*. The most important functions of this category
150 are those converting the relative humidity to the mixing ratio (`rh2w`), the
151 relative humidity to the specific humidity (`rh2shum`), the mixing ratio to
152 the dew point temperature (`w2Td`), and the mixing ratio to the specific
153 humidity or the reverse (`w2q` and `q2w`), or `e2w` that to convert the partial
154 pressure of water vapour to the mixing ratio.

155 • Saturation mixing ratios or pressures

156 This class includes all functions that use the Clausius-Clapeyron equation
157 to calculate the saturation mixing ratios or pressures. The most impor-
158 tant function is `saturation_pressure_H20`, which computes the satura-
159 tion pressure e_s in Pa as a function of the temperature. As well as this,
160 `saturation_mixing_ratio` returns the saturation mixing ratio w_s in kg
161 kg^{-1} .

162 • State and evolution of an air parcel

163 The package calculates the internal state of a parcel using a given pressure,
164 temperature, and mixing ratio with the function `parcelState`. However,
165 to calculate the vertical evolution of an air parcel, *aiRthermo* determines
166 the correct function depending on the state of the particle and the envi-
167 ronment. General ascent from a given initial pressure to the final pressure
168 is computed by the `adiabatic_ascent` function, which selects the type of
169 evolution depending on the saturation of the parcel. Downwards evolution
170 can also be computed using `AnyAdiabaticDown`, but it requires the initial
171 amount of water available in the cloud for evaporation (in kg/kg). Con-
172 versions between the potential temperature, Temperature, and pressure
173 can be calculated using `PT2Theta`, `PTheta2T`, and `TTheta2P`, which are
174 useful if dry adiabatic processes are occurring.

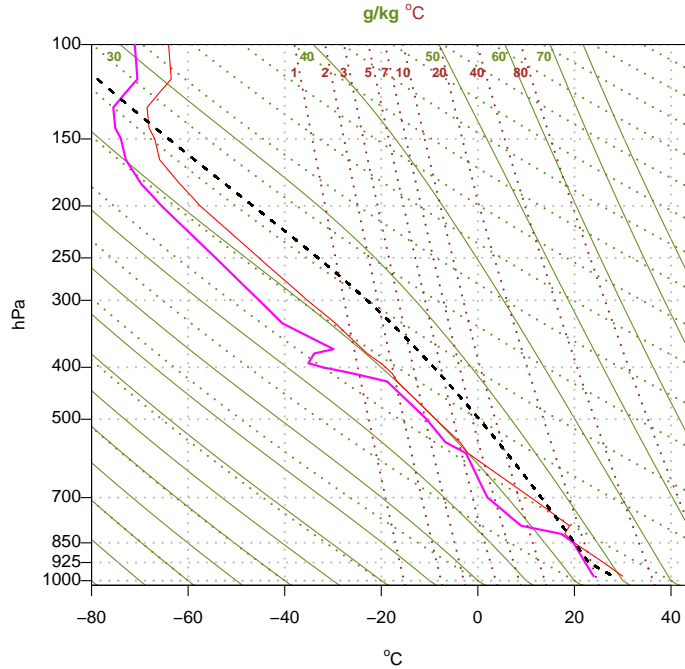


Figure 2: Example of a Stüve diagram plotted with *aiRthermo*. The temperature (red) and dew point temperature (magenta) of the sounding area are plotted together with the evolution of the lifted air parcel (shaded black line).

175 • **Instability indices**

176 Several functions compute common instability indices such as K, the Total-
 177 Totals, the Showalter, and the LI indices (functions `Kindex`, `TTindex`,
 178 `Sindex`, and `LIindex`). `CAPE_CIN` calculates the values of CAPE and
 179 CIN, the LCL, the Level of Free Convection (LFC), the End Level (EL),
 180 and the trajectory, followed by the lifted parcel. The `PlowTop` argument
 181 provides the width of a slab across which a vertical average will be taken
 182 through the bottom of the sounding to obtain the initial conditions of
 183 the ascending parcel. `precoolType` determines the type of precooling that
 184 must be applied to the initial parcel. `upToTop` controls whether the lifted
 185 particle continues upwards after it first crosses the ambient sounding.

186 • **Stüve diagrams**

187 The `stuve_diagram` function is included to allow the creation of high-

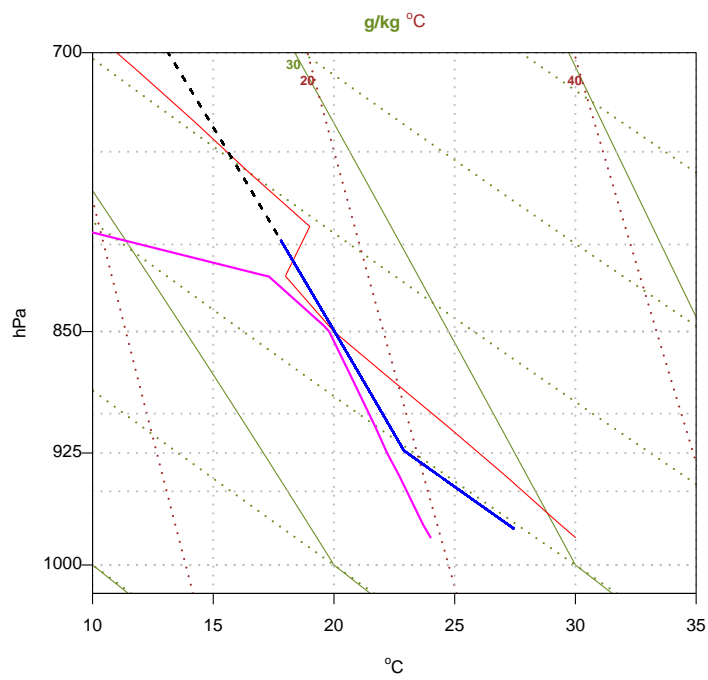


Figure 3: Detail of the example Stüve diagram in Figure 2 magnifying its lower levels to show the different evolutions of the lifted parcel depending on the value of the *upToTop* parameter (*TRUE* (black line) and *FALSE* (blue line)).

188 quality Stüve diagrams. As well as generating the Stüve diagram, they
189 can also represent the trajectory followed by any ascending air parcel
190 (when `CAPE_CIN` is called using `getLiftedBack=TRUE`).

191 Figure 2 shows the Stüve diagram corresponding to the Davenport sample
192 sounding. Major differences in the estimated CAPE for the sounding can appear
193 if the `upToTop` attribute is set to `TRUE` or `FALSE` (discontinuous black and
194 blue lines in Figure 3 respectively) as the lifting particle slightly crosses the
195 sounding at low levels. These results can be controlled using the appropriate
196 parameters. Sensible default parameters are used by the function if they are
197 not explicitly provided by the user.

198 Listing 1 shows the manner in which `CAPE_CIN` can be used to produce a
199 figure similar to Figure 2. Each time `CAPE_CIN` is called, `getLiftedBack` can be
200 set to `TRUE` so that the trajectory of the lifted parcel is returned to the calling
201 environment. The discontinuous black line in Figure 2 shows the trajectory of
202 the ascending parcel when it is requested to ascend to the top of the sounding
203 (`upToTop=TRUE`). If `FALSE` is assigned to the `upToTop` parameter, the parcel
204 (blue line in Figure 3) stops after the first time that the ascending parcel is
205 not buoyant, thus leading to a severe underestimation of the value of CAPE.
206 The buoyancy is evaluated using virtual temperatures for the parcel and the
207 environment in all cases.

Listing 1: Evaluation of CAPE (CIN) and a representation of a sounding

```
208 data(RadiosondeDavenport)  
209 dPs<-RadiosondeDavenport[,1]*100  
210 dTs<-C2K(RadiosondeDavenport[,3])  
211 dws<-RadiosondeDavenport[,6]/1000  
212 # Dew point T in Celsius for plotting  
213 dTd<-RadiosondeDavenport[,4]  
214 # Initial conditions are known.  
215 # upToTop<-FALSE "stop after the parcel is not buoyant"  
216 capeOut<-CAPE_CIN(Ps=dPs, Ts=dTs, ws=dws, deltaP=1,  
217                   P0=97500, T0=300.6, w0=0.01936, upToTop=FALSE,  
218                   getLiftedBack=FALSE, precoolType="none")  
219
```

```

220 # some of the information from the output object
221 print(paste("Davenport_CAPE:" , capeOut$cape ,
222           " J/kg" , "CIN:" , capeOut$cin , " J/kg" , "LFC" ,
223           capeOut$apLFC$P , "Pa" , capeOut$apLFC$Temp , "K" ))
224 # Whole sounding (upToTop<-TRUE), parcel is returned
225 capeOut<-CAPE_CIN(Ps=dPs , Ts=dTs , ws=dws , deltaP=1 ,
226                 P0=97500 , T0=300.6 , w0=0.01936 ,
227                 upToTop=TRUE , getLiftedBack=TRUE)
228 # Plot sounding
229 plot<-stuve_diagram( Pres=dPs/100 , Temp=K2C(dTs))
230 # Dew point temperature
231 lines(dTd , dPs/100 , col="magenta" , lwd=2)
232 # Lifted parcel (upToTop=TRUE)
233 lines(K2C(capeOut$Tl) , (capeOut$P1/100) ,
234       col="black" , lwd=2 , lty=2)
235

```

236 A profiler has been used to evaluate the CPU time that *aiRthermo* requires
237 to calculate saturated adiabatic evolution from an initial pressure of 950 hPa
238 to 200 hPa (pressure step of 1 Pa). A routine that performs the same compu-
239 tation using an R iteration was also written. The evolution takes 50 ms using
240 *aiRthermo*, however, that using R alone takes approximately 1500 ms. There-
241 fore, the speed is approximately thirty times faster due to the use of the C core
242 for computing vertical evolutions.

243 5. Case study: Instability indices over the Iberian Peninsula (2010- 244 2014)

245 The performance of the package has been checked using a large number of
246 real soundings measured over the Iberian Peninsula (Figure 1), an area well
247 known for the development of convective systems [22].

248 In this case study, 24072 soundings from the eight stations were processed to
249 estimate the performance of the package using realistic data. CAPE and CIN
250 were computed using a very low vertical step (0.5 Pa), and all the soundings
251 extended to the top of the sounding. CAPE and CIN were computed twice for

252 two different initial parcels (one from the lowest point of the sounding, and the
253 second from a low-level average). The K, Total-Totals, Lifted, and Showalter
254 indices were also computed. This process took six hours on a common desktop
255 computer (2015) running Linux.

256 The results of CAPE computed by *aiRthermo* and the value stored at the
257 University of Wyoming server were compared. Figure 4 shows the resulting
258 scatterplots for A Coruña (Atlantic site) and Barcelona (Mediterranean). The
259 Pearson’s correlation coefficient R value is very good in both cases, at 0.94 in A
260 Coruña (3521 soundings) and 0.98 in Barcelona (3575 soundings). The values
261 of R were within a range of $[0.94, 0.99]$ at all stations.

262 As shown in Figure 4, there are some clear outliers in the scatterplots. An
263 analysis comparing some of these points (tagged with numbers) that yield dif-
264 ferent CAPE values to those provided by Wyoming has been conducted. In the
265 first case (Barcelona, May 8th, 2011), *aiRthermo* computes a CAPE of 2625
266 J/kg, while Wyoming computes a CAPE of 1155 J/kg. This difference is due to
267 the sensitivity to the initial conditions that characterise the CAPE value. We
268 cannot be certain of the manner in which the vertical evolutions are initialised
269 in the Wyoming server, but we have verified that errors smaller than 2% in
270 the estimation of the initial state of the parcel (its pressure or temperature)
271 lead to differences of 100% between the resulting CAPE values of this sounding,
272 characterised by almost complete saturation until 500 hPa. The second case
273 (Barcelona, August 26th, 2012) corresponds to a sounding with no moisture in
274 the mixing ratio column of the Wyoming archive beyond 700 hPa. As the data
275 are missing, *aiRthermo* is forced to stop at that level because there is no mois-
276 ture information in the mixing ratio column. If the mixing ratio is assumed to
277 be zero beyond that point and the whole sounding is processed, the differences
278 are small (1767 J/kg in *aiRthermo* vs 1651 J/kg in Wyoming). Finally, the third
279 analysed example corresponds to A Coruña, 27th June 2012. In this case, the
280 mixing ratio column in the table distributed by Wyoming contains zeros. How-
281 ever, the moisture is not completely missing. A value of CAPE similar to that
282 on Wyoming’s web page (1247 J/kg) is computed by *aiRthermo* (1211 J/kg) if
283 the moisture is computed from the dew point temperature in Wyoming’s server.

284 Although the exact initial conditions used for computing CAPE in Wyoming

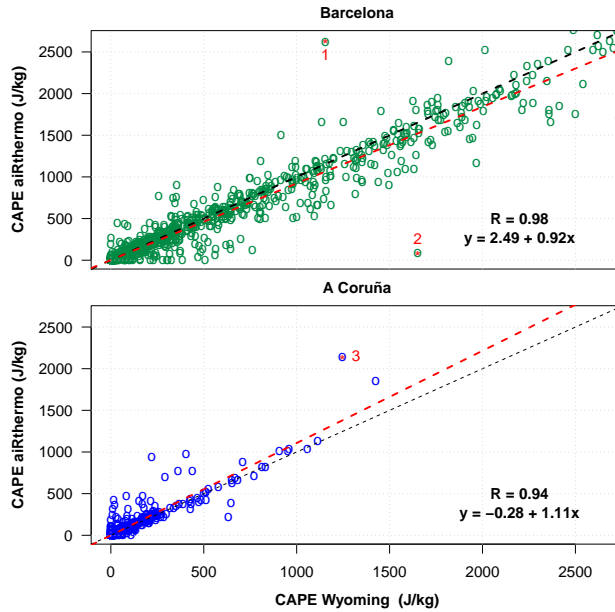


Figure 4: Scatterplot of the CAPE values computed by *aiRthermo* and the corresponding values computed by the University of Wyoming for Barcelona (top) and A Coruña (bottom).

285 are unknown to us, which means that the sensitivity of the results caused by
 286 this cannot be fully evaluated, the least squares regression lines exhibit high
 287 agreement, with slopes very close to one (0.92 and 1.11, as shown in Figure 4).

288 6. Discussion and Conclusions

289 This paper presents a new package for R, *aiRthermo*, which is available in the
 290 open-source repository for R packages CRAN. It provides R with new functions
 291 in the field of atmospheric thermodynamics. These capabilities considerably
 292 extend the analyses that can be conducted from inside the R interpreter.

293 In the field of storm forecasting, the parameters that can be computed using
 294 *aiRthermo* have been used to forecast storms in Belgrado [23]. Therefore, the
 295 ability to run different statistical models and verification procedures from inside
 296 R alongside computing the indices themselves could boost this type of study.
 297 Similar analysis [24] has been conducted in the Arctic region (Bjørnøya, Jan
 298 Mayen and Svalbard Islands), and the results of the distribution of instability
 299 indices were very different to those of other European regions [25]. The instabil-

300 ity indices available in *aiRthermo* have been combined with satellite data from
301 regional instability indices in Africa [26] and India [27]. These results show
302 that analysing the climatological distribution of atmospheric instability [25], its
303 interannual variability, and its expected future distribution [22] under globally
304 changing conditions can benefit from using *aiRthermo*. As some instability in-
305 dices must be computed from vertical adiabatic evolutions of air parcels close
306 to the surface (the case of CAPE or LI), analysing these distributions is eas-
307 ier if every sounding can be processed in a much shorter time (as performed
308 by *aiRthermo*). This will increase the use of instability indices in downscaling
309 methods that are designed to operate on daily precipitation [28], even for long
310 periods of time.

311 As well as precipitation forecasts, other damaging effects can be derived from
312 atmospheric convective instability. Atmospheric instability indices are used to
313 analyse lightning in Western Patagonia [29] and the Iberian Peninsula [30]. They
314 also serve as prognostic variables when statistically downscaling wind variabil-
315 ity [31]. The aircraft safety field is another area where the thermodynamic
316 properties of air, particularly when it is close to saturation, are important, both
317 at the surface, during the prevention of fog (when studying ice fog, for exam-
318 ple [32]), or aloft to prevent the aircraft from icing [33].

319 The authors have also recently used the package for educational purposes in
320 the case of M.Sc. studies in the University of Basque Country [34]. Students
321 use the routines in the package to numerically simulate the vertical evolution of
322 parcels and verifying their results with those they obtain from thermodynamic
323 diagrams. They also assess the role of large-scale moisture convergence against
324 convection in intense precipitation events by computing instability indices for
325 air parcels, and solve simple exercises related to the Föhn effect [13].

326 There is a perpetual need for verifying complex forecasting systems based
327 on advanced numerical models. The diagnostics available in *aiRthermo* are of-
328 ten used in these verification processes [35]. The direct availability of these
329 diagnostics from the package used for the verification (frequently R itself) will
330 quicken the development of these models. This also includes new satellite sys-
331 tems, such as GOES-R [36, 37], which can produce integrated instability indices,
332 such as LI, almost in real-time. The ability to easily compute the values of LI,

333 CAPE, or other indices from soundings, satellite sounders [38, 39], and numer-
334 ical model results will increase the ease of the interoperability of modelled and
335 observational data (soundings, satellite-derived products, and remotely sensed
336 information) in newly developed operational environmental forecasting systems.

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343 versity of Wyoming, Dept. of Atmospheric Science, are greatly acknowledged.
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345 proved our manuscript.

346 8. Software and data availability

347 The software presented in this paper is a package prepared to work within
348 the R data analysis suite and was developed by the authors:

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362 Year first available in CRAN: 2017.

363 Hardware required: It has been tested on laptops, Desktops, and worksta-
364 tions running Mac OS, Windows, and Linux.

365 Availability: The software and datasets are freely available (GPL-3 license)
366 in the Comprehensive R Archive Network (CRAN):

367 <https://cran.r-project.org/package=aiRthermo>

368 The package can be installed from any of the mirrors, which is usual for R
369 packages, by typing `install.packages("aiRthermo")` into the R interpreter.
370 The size of the package ranges from 400 to 500 Kb, depending on whether the
371 source or Windows-compiled version is downloaded. The CRAN servers allow
372 anonymous access to the package. The software is written in R and C. The
373 manual is also provided at the CRAN server [40].

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