

Temperature-dependence of the clear-sky feedback in radiative-convective equilibrium

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Key Points:

- The temperature-dependence of the clear-sky radiative feedback saturates at $-1.2 \text{ W m}^{-2} \text{ K}^{-1}$
- Masking effects by water-vapor at the flanks of the CO_2 band weaken the radiative forcing at high column water vapor
- Barring insolation changes Earth's climate is stable to even unimaginably large increases in atmospheric CO_2 .

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Abstract

We quantify the temperature-dependence of the clear-sky climate sensitivity in a one-dimensional radiative-convective equilibrium model. The atmosphere is adjusted to fixed surface temperatures between 280 K and 320 K while preserving other boundary conditions in particular the relative humidity and the CO₂ concentration. We show that an out-of-bounds usage of the radiation scheme RRTMG can lead to an erroneous decrease of the feedback parameter and an associated “bump” in climate sensitivity as found in other modelling studies. Using a line-by-line radiative transfer model, we find an almost constant longwave radiative feedback at surface temperatures above 300 K. However, the line-by-line simulations also show a slight decrease in climate sensitivity when surface temperatures exceed 310 K. This decrease is caused by water-vapor masking the radiative forcing at the flanks of the CO₂ absorption band, which reduces the total radiative forcing by about 18 %.

Plain Language Summary

The climate feedback parameter describes how the net radiative balance at the top of the atmosphere changes with surface temperature. The magnitude of the feedback parameter here depends on the current state of the climate system. For example, a warmer climate state is accompanied by a moister atmosphere which limits the climate feedback and hence increase climate sensitivity – which is the surface warming due to a doubling of CO₂. Other modelling studies have shown that the climate sensitivity will first increase in a warmer reference climate, but decrease again when surface temperatures exceed 310 K. In this study, we are using a reference radiative transfer model to show how the misuse of a simplified radiation scheme leads to this spurious signal in the estimation of the climate feedback parameter. In addition, we explain how changes in the H₂O and CO₂ concentrations influence the spectral distribution of both the feedback parameter and the radiative forcing.

1 Introduction

The state-dependence of the climate sensitivity is of great interest when studying climate change as it influences the interpretation of the proxy record (Manabe & Bryan, 1985; Kutzbach et al., 2013), historical temperature observations (Andrews, 2014; Gregory & Andrews, 2016), and the interpretation of differences among models (Bourdin

46 et al., 2021). While many studies focus on cloud feedbacks due to changes in self-aggregation,
 47 cloud amount, or cloud height (Becker & Wing, 2020; Zelinka et al., 2020; Bony et al.,
 48 2016), there is a growing but still inconclusive literature on the seemingly simpler ques-
 49 tion of the clear-sky radiant response to warming.

50 Recent modelling studies, ranging from conceptual (Meraner et al., 2013), to cloud-
 51 resolving, models (Romps, 2020), find that after an initial decrease the magnitude of the
 52 clear-sky feedback parameter, λ , again increases at yet higher surface temperatures (T_s).
 53 This non-monotonicity manifests itself as a pronounced “bump”, a maximum in the clear-
 54 sky climate sensitivity, \mathcal{S} , at $T_s \approx 310$ K. Seeley and Jeevanjee (2021) describe a phys-
 55 ical mechanism that explains the changing temperature-dependence of λ : when the rise
 56 of the temperature is tied to the rise of CO_2 , the increased CO_2 broadens the spectral
 57 interval over which CO_2 is the dominant absorber, thereby coupling OLR in these spec-
 58 tral regions to the tropospheric temperature, and hence T_s in a way that leads to a more
 59 negative λ with warming. The work by Seeley and Jeevanjee (2021) provides an elegant
 60 physical explanation for the climate sensitivity “bump” in studies with varying CO_2 con-
 61 centration (e.g. Romps, 2020) and in doing so shows how λ effectively depends on CO_2 .
 62 However, their mechanism fails to explain a similar “bump” in \mathcal{S} as temperature increases
 63 in constant- CO_2 simulations as in Meraner et al. (2013). Moreover, coupling tempera-
 64 ture changes to CO_2 , while physical, makes it difficult to separate the state-dependence
 65 of λ on T_s from its dependence on CO_2 .

66 In this study, we calculate \mathcal{S} as a function of a fixed T_s , for $T_s \in [280 \text{ K}, 320 \text{ K}]$.
 67 After the atmosphere has equilibrated to the boundary conditions and the chosen T_s , the
 68 radiative feedback is computed as the change in OLR between simulations at increas-
 69 ing T_s (Section 2). Calculations were initially performed using a fast radiative transfer
 70 model (Mlawer et al., 1997), identical to that used in many climate modelling studies.
 71 To check the calculations of the more parameterized fast radiative transfer model, and
 72 to understand how the spectral forcing and feedback associated with a doubling of at-
 73 mospheric CO_2 depends on temperature, we also perform calculations with a line-by-line
 74 model (Section 2). Our approach differs from earlier studies (Goldblatt et al., 2013; Klufft
 75 et al., 2019) in that the temperature profile and heating rates are allowed to interact.
 76 We find that qualitative errors from the fast radiative model become pronounced as T_s
 77 increases above 300 K, and it over estimates the temperature-dependence of \mathcal{S} by more
 78 than a factor of two as compared to the line-by-line model reference (Section 3).

79 Studies of the clear-sky feedback date back to Simpson (1928), who proposed that
 80 – in an atmosphere whose optical properties arise from a condensible species (water) –
 81 OLR decouples from T_s when the atmosphere becomes optically thick. Ingram (2010)
 82 brought these ideas to the attention of the climate community (in the meantime plan-
 83 etary scientists, initially unaware of Simpson’s work, had come to similar conclusions)
 84 and concluded that if the water vapor concentration is a function of temperature only,
 85 a warming atmosphere will increase its optical thickness (and hence its emission height)
 86 in a way to maintain a constant emission temperature. For Earth’s atmosphere this hap-
 87 pens when $T_s > 300$ K (Goldblatt et al., 2013; Koll & Cronin, 2018). This decoupling
 88 was later (and independently) shown to underpin a limit to how much energy Earth’s
 89 troposphere can radiate to space in the thermal infrared (Nakajima et al., 1992), with
 90 runaway (greenhouse) warming ensuing when the absorbed insolation exceeds this limit
 91 (Kasting, 1988; Nakajima et al., 1992; Goldblatt et al., 2013, 2017). These findings en-
 92 courage the expectation that λ and hence \mathcal{S} will increase monotonically with T_s , increas-
 93 ingly so for $T_s > 310$ K, rather than to first increase and then decrease, as found by Meraner
 94 et al. (2013). Our line-by-line calculations show a different behavior: λ asymptotes to
 95 a constant, but negative value, with warming. To understand this behavior we quantify
 96 the magnitude of the radiative forcing ΔF from a CO_2 doubling. At moderate temper-
 97 atures we find an increase in ΔF with surface warming, which is driven by higher emis-
 98 sion temperatures in the center of the CO_2 band due to a deepening of the troposphere
 99 (Huang et al., 2016). At warm temperatures this trend is reversed and ΔF decreases.
 100 Together this leads to a reduction of \mathcal{S} with warming. By considering the spectral re-
 101 sponse to warming and forcing (Section 4) we are able to understand this behavior, also
 102 in light of the earlier literature.

103 2 Methods and data

104 To analyze how the clear-sky climate sensitivity, \mathcal{S} , varies with surface tempera-
 105 ture, T_s , we use the one-dimensional radiative convective model konrad (Dacie et al., 2019;
 106 Kluft et al., 2019) equilibrated at prescribed values of T_s between 280 K to 320 K with
 107 a fixed relative humidity of 80 % (see Appendix for a more detailed model description).
 108 Figure 1a shows the resulting temperature profiles as a function of atmospheric pressure
 109 p .

The RCE simulations are performed for CO₂ concentrations of 348 ppmv and 696 ppmv which allows us to compute the radiative forcing ΔF and the feedback parameter λ . We define ΔF at a given T_s as the difference in net radiation balance ΔN at the top of the atmosphere between these two CO₂ concentrations

$$\Delta F = \Delta N_{696 \text{ ppmv}} - \Delta N_{348 \text{ ppmv}} \quad (1)$$

The feedback parameter λ is defined as the change in ΔN between simulations at constant CO₂ = 348 ppmv and different T_s

$$\lambda(T_s) = \frac{\Delta N(T_s + \Delta T) - \Delta N(T_s - \Delta T)}{2\Delta T} \quad (2)$$

110 with surface temperature difference $\Delta T = 1$ K. With this approach, we can study the
 111 temperature-dependence of the radiative forcing ΔF , the climate feedback λ , and the
 112 resulting climate sensitivity $\mathcal{S} = -\Delta F/\lambda$.

113 To check our method we have also performed simulations with a coupled T_s and
 114 computed λ as the regression of ΔN over ΔT_s during a perturbed simulation. We find
 115 that the results are in very good agreement with those obtained using Equation 2. How-
 116 ever, the strong temperature-dependence of λ makes the linear regression error-prone,
 117 which mostly manifests in spurious signals in the estimated effective forcing (y -intercept
 118 of the regression). Therefore, we opted for the well-established fixed SST approach.

119 Baseline simulations are performed using the Rapid Radiative Transfer Model for
 120 GCMs (RRTMG, Mlawer et al., 1997). RRTMG is a fast radiation scheme which uses
 121 the correlated- k method with precalculated lookup tables for computational efficiency.
 122 For line-by-line simulations we replace the RRTMG longwave radiative transfer calcu-
 123 lations with calculations using the Atmospheric Radiative Transfer Simulator (ARTS,
 124 Eriksson et al., 2011; Buehler et al., 2018). ARTS represents the longwave radiative fluxes
 125 based on 32 768 equidistant frequency points (lines) between 10 cm^{-1} and 3250 cm^{-1} ($\Delta\nu =$
 126 0.1 cm^{-1}). Explicitly resolving the spectrum of OLR later allows us to investigate con-
 127 ceptual ideas about the dependence of OLR on T_s in different spectral regions.

128 For the sake of simplicity and to facilitate comparisons with previous modelling stud-
 129 ies we do not consider the effects of ozone. We have performed calculations in which ozone
 130 is allowed to change, and while the basic physics that we describe are not influenced by
 131 this elaboration, as λ becomes small the effect of ozone can become important. Quan-
 132 titatively its influence is found to depend on the details of its representation, particu-

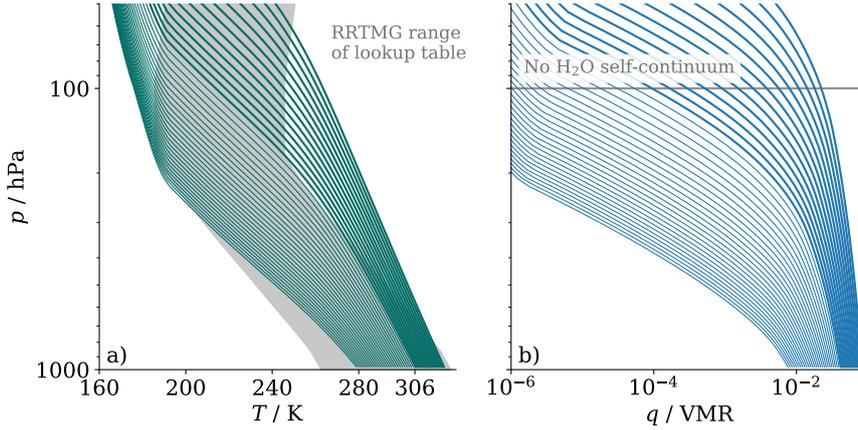


Figure 1. Equilibrium temperature *a)* and water-vapor volume mixing ratio *b)* profiles at different surface temperatures but constant CO₂ concentrations as a function of atmospheric pressure. The figure is clipped at 50 hPa to better visualize the troposphere. In addition, the RRTMG reference temperature range is shown as grey area.

133 larly in light of the deepening of the troposphere with warming, an interesting issue that
 134 we are beginning to explore together with experts on ozone chemistry.

135 The representation of the climate system, or even just the climate of the tropics,
 136 in terms of cloud (as well as aerosol and ozone) free radiative-convective equilibrium (RCE)
 137 is a strong, but common, simplification. The Charney (1979) report took it as a start-
 138 ing point and a large literature since then has found RCE solutions to be informative
 139 of how different physical processes influence climate sensitivity. For this reason RCE re-
 140 mains a well studied model problem (Popke et al., 2013; Stevens & Bony, 2013; Wing
 141 et al., 2017; Bourdin et al., 2021), one which for reasons elegantly articulated by Polya
 142 (1962), is worth first understanding.

143 Further information about konrad’s configuration, RRTMG, and ARTS is given in
 144 the Appendix.

145 **3 Temperature-dependence of the feedback parameter λ**

We run konrad for T_s between 280 K and 320 K to quantify the temperature-dependence of the feedback parameter

$$\lambda = f(T_s; I, \alpha, \text{RH}, \chi) \quad (3)$$

146 with constant values of insolation I , surface albedo α , relative humidity RH, and the gaseous
 147 composition χ . Seeley and Jeevanjee (2021) consider the related problem $\lambda = f(T_s, \text{CO}_2; \dots)$,
 148 where the CO_2 concentration is variable.

149 For low temperatures, calculations based on RRTMG and ARTS agree well with
 150 one another. Figure 2 shows the radiative forcing ΔF and the feedback parameter λ (as
 151 defined in Section 2), as well as the resulting equilibrium climate sensitivity \mathcal{S} , as a func-
 152 tion of T_s . Both results using RRTMG (grey), and the line-by-line radiative transfer model
 153 ARTS (green), show that λ (Figure 2b) increases from $-2.1 \text{ W m}^{-2} \text{ K}^{-1}$ to $-1.3 \text{ W m}^{-2} \text{ K}^{-1}$
 154 as T_s increases from 280 K to 300 K. A more detailed feedback analysis (not shown) iden-
 155 tifies this increase with the temperature-dependence of the water-vapor feedback.

156 For $T_s > 300 \text{ K}$, calculations with RRTMG result in a pronounced local maximum,
 157 or “bump”, in \mathcal{S} . This is seen in Fig. 2, where \mathcal{S} increases from less than 3 K at $T_s =$
 158 300 K to about 8 K at $T_s = 320 \text{ K}$, and then rapidly decreases to less than 2 K at $T_s =$
 159 320 K. Fig. 2 further show that RRTMG’s response can be attributed changes of the feed-
 160 back parameter $\partial_{T_s} \lambda$, rather than the forcing. Hence the bump, and its origins, are sim-
 161 ilar to what was found in other studies (Meraner et al., 2013; Romps, 2020) using correlated-
 162 k radiative transfer. When using ARTS, however, $\partial_{T_s} \lambda$ does not increase. In contrast,
 163 the temperature-dependence of \mathcal{S} begins to decrease at 305 K and λ converges to an al-
 164 most constant value of $-1.2 \text{ W m}^{-2} \text{ K}^{-1}$.

165 RRTMG, and other fast-radiative transfer schemes, aggregate absorption features
 166 into bands, within which optical properties are calculated by interpolating across pre-
 167 computed look-up tables. This reduces the computational intensity and speeds up the
 168 calculations many fold. In RRTMG the look up tables are based on an assumed atmo-
 169 spheric composition and thermal structure, close to those of the present-day Earth (Mlawer
 170 et al., 1997, their Sec. 3.2). As it turns out, how one interprets the word ‘close’ can be
 171 problematic. For instance, while RRTMG is documented to be valid for T_s as high as
 172 320 K, this is based on a temperature profile representative of mid-latitude summer and
 173 assumes that there are no changes in the temperature lapse rate with increasing surface
 174 temperature (Mlawer et al., 1997, their Sec. 3.2). As a consequence, the temperature lapse-
 175 rate in the lookup table is larger than the moist-adiabat, which implies mid- and upper-
 176 tropospheric temperatures that are out of bounds at T_s above 306 K (see Figure 1). Popp
 177 et al. (2015) attempted to minimize the resultant errors by clipping the temperatures

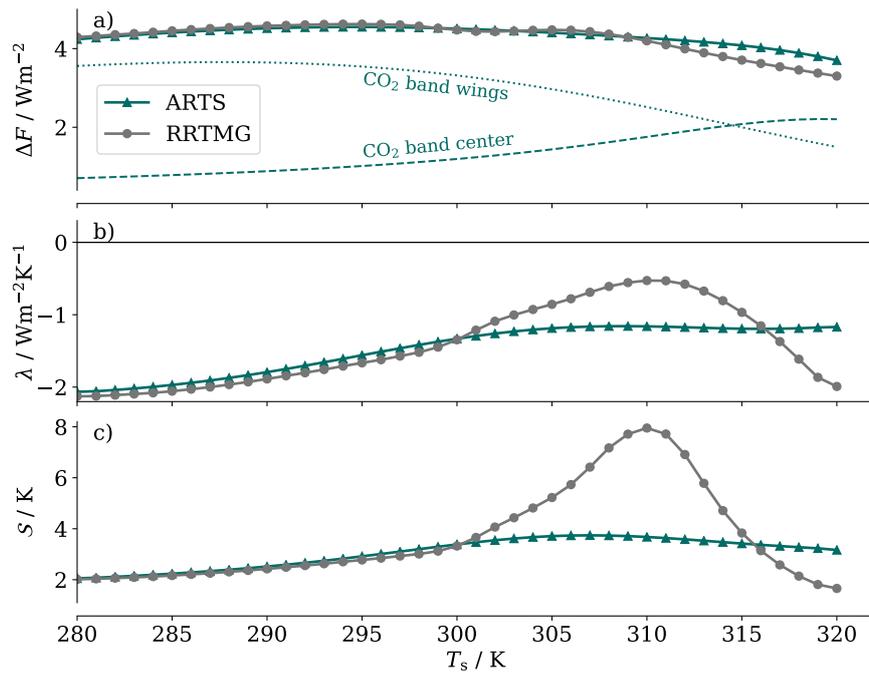


Figure 2. *a)* Effective radiative forcing ΔF , *b)* climate feedback parameter λ , and *c)* equilibrium climate sensitivity S as function of surface temperature T_s . All quantities are shown for experiments using the radiation scheme RRTMG (grey) and the line-by-line radiative transfer model ARTS (green).

178 to acceptable bounds when performing the gaseous look-up. The look-up tables are only
 179 one source of error. Another, which we identified, arises from RRTMG’s calculation of
 180 the water-vapor self continuum. For computational expediency this is fit to only two ref-
 181 erence values (at 260 K and 296 K) and is neglected entirely for pressures less than 100 hPa.
 182 This procedure is error prone as T_s increases above 296 K, the tropopause deepens, and
 183 when the window closes (which depends on relative humidity). For the case of RRTMG,
 184 these errors lead to an overestimation of OLR, which is misinterpreted as a decrease (more
 185 negative) of λ at high T_s . Coincidentally, this happens around the same temperature range
 186 at which the CO₂ mechanism described by Seeley and Jeevanjee (2021) begins to work.

187 In conclusion, using a line-by-line radiation model we find a robust increase of λ
 188 for T_s up to 305 K. Errors in the calculation of longwave irradiances by RRTMG are shown
 189 to be the cause of a spurious “bump” in \mathcal{S} . This “bump” looks similar, but is entirely
 190 unrelated, to the local maximum in \mathcal{S} that Seeley and Jeevanjee (2021), find (and phys-
 191 ically explain), when CO₂ is allowed to covary with T_s . For fixed CO₂, as T_s increases,
 192 λ asymptotes to a near-constant value of around $-1.2 \text{ W m}^{-2} \text{ K}^{-1}$. That the feedback
 193 becomes asymptotically constant as the window closes is consistent with expectations
 194 from the literature on clear-sky radiative feedbacks in moist atmospheres (Ingram, 2010;
 195 Goldblatt et al., 2017). Its substantially negative value was less expected and has pro-
 196 found implications, something we address in more detail in the following section.

197 **4 Spectral analysis of λ and ΔF**

198 To understand why λ doesn’t reduce to zero with warming, as one might expect
 199 based on a consideration of the response of water vapor alone to warming, we here ex-
 200 amine the spectral feedback parameter λ_ν . This framework was used by Kluft et al. (2019)
 201 as well as Seeley and Jeevanjee (2021), and can also be used to study the role of differ-
 202 ent spectral regions in changing ΔF and \mathcal{S} . The important difference between our sit-
 203 uation, and the situation envisioned by Simpson (1928), is that H₂O is not the only ab-
 204 sorber in the infrared. Were that the case it would not be possible to force the system
 205 by increasing atmospheric concentrations of CO₂. The problem as we pose it here, is not
 206 how Earth can respond to energy accumulated by an external process, such as insola-
 207 tion or accretion of extra-planetary material (Abe & Matsui, 1988; Kasting, 1988; Naka-
 208 jima et al., 1992), but rather how the reduction of infrared irradiance of the atmosphere
 209 can be compensated through warming.

210 The spectral feedback parameter λ_ν can be derived from our line-by-line calcula-
 211 tions using Eq. (2). Figure 3b shows the smoothed λ_ν as a function of wavenumber ν
 212 for simulations at different temperatures (and hence absolute humidity). There is a strong
 213 temperature-dependence of λ_ν in the atmospheric window between 800 cm^{-1} and 1200 cm^{-1} .
 214 This is driven by the increasing water vapor concentration in the warming troposphere,
 215 as λ_ν is indeed close to zero as soon as the atmosphere becomes fully opaque at high tem-
 216 peratures (darker blue shades) and stays close to zero for higher T_s . Hence, our results
 217 link the findings of Koll and Cronin (2018) with the studies by Nakajima et al. (1992)
 218 and Goldblatt et al. (2013).

219 In our simulations the total λ remains negative definite for all T_s . The thermal Planck
 220 feedback in the CO_2 bands around 667 cm^{-1} maintain a stable feedback with $\lambda \approx -1.2\text{ W m}^{-2}\text{ K}^{-1}$.
 221 Adopting the analogy introduced by Seeley and Jeevanjee (2021), the infrared emission
 222 attributable to tropospheric CO_2 acts as a spectral radiator fin, stabilizing the climate
 223 to greenhouse forcing. This explains why a runaway greenhouse effect can not arise from
 224 the effect of CO_2 on thermal emission alone. The same mechanism which allows an in-
 225 crease in CO_2 , or any other temperature independent greenhouse gas, to increase the ra-
 226 diative forcing, will also increase the radiative feedback of the system. Even in a con-
 227 stant CO_2 scenario, a moistening of the atmosphere will only dampen the thermal Planck
 228 feedback but never eliminate it. Ingram (2010) speculated that this might be the case,
 229 here we elaborate this thought and show that the ability of the atmosphere to maintain
 230 a feedback in some parts of the spectrum is intrinsically tied to the existence of a radia-
 231 tive forcing. In other words, if the system can be radiatively forced by a temperature-
 232 invariant greenhouse gas, i.e. CO_2 , it also has the ability to maintain a stable climate.

233 Despite the constant value of λ , the magnitude of \mathcal{S} decreases at $T_s > 310\text{ K}$. This
 234 decrease in \mathcal{S} is driven by a decrease of ΔF with warming. Usually, the radiative forc-
 235 ing is thought to increase with T_s (Huang et al., 2016). Such an effect is apparent in our
 236 simulations, but only for lower values of T_s , up to 300 K (Figure 2a). This strengthen-
 237 ing of ΔF with warming arises from a larger contribution from the band center (between
 238 620 cm^{-1} to 700 cm^{-1}). At higher T_s , ΔF decreases, so that with $T_s = 320\text{ K}$ it is 18 %
 239 less than its value at 295 K . Fig 3 shows that the reduction in ΔF with warming is due
 240 to a weakening contribution from the edges of the 667 cm^{-1} CO_2 band. At $T_s = 280\text{ K}$,
 241 15 % of the forcing is carried by the band center, at 320 K the forcing from the band cen-

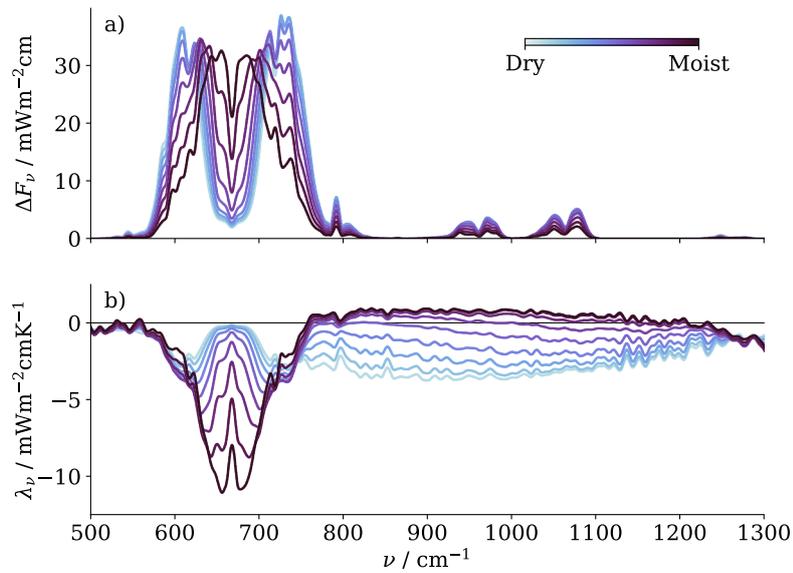


Figure 3. *a)* Spectral radiative forcing F_ν and *b)* spectral feedback parameter λ_ν as functions of wavenumber ν . Darker shades of blue represent a warmer and moister atmosphere. The spectra are smoothed using a 15 cm^{-1} running mean and zoomed to a range of 500 cm^{-1} to 1300 cm^{-1} to better resolve the CO_2 absorption band around 660 cm^{-1} and the atmospheric emission window between 800 cm^{-1} to 1200 cm^{-1} . The actual line-by-line simulations cover a wavenumber range from 10 cm^{-1} to 3250 cm^{-1} .

242 ter has increased more than three-fold and is responsible for 60% of the total forcing (Fig-
 243 ure 2a).

244 CO₂ absorption is so strong near the central absorption feature, that emission to
 245 space from these wavelengths originate in the stratosphere. Only lines whose emission
 246 height resides in the troposphere – where temperatures decrease with height – contribute
 247 to reduced emissions, and hence forcing from increasing CO₂ concentrations. As the tropopause
 248 rises with warming, an increasing fraction of the OLR originates from CO₂ in the tro-
 249 posphere, and its changes can contribute to the forcing. As increasing water vapor closes
 250 the window at $T_s > 300$ K, emission by H₂O increasingly dominates over emission by
 251 CO₂ on the flanks of the CO₂ band. This reduces the contribution of tropospheric CO₂
 252 to the OLR, thereby reducing the contribution of its changes to forcing. The latter in-
 253 creasingly dominates at warmer temperatures, weakening ΔF from a doubling of CO₂
 254 by about 18% (from a value around 4.5 W m^{-2} to 3.7 W m^{-2}), consistent with an an-
 255 alytical model of the CO₂ forcing by Jeevanjee et al. (2020).

256 Seeley and Jeevanjee (2021) demonstrated how an increase in CO₂ concentration
 257 strengthens the CO₂ absorption band in the atmospheric window: at some point the CO₂
 258 replaces H₂O as the dominant absorber and acts as a “CO₂ radiator fin”. To understand
 259 the effects of warming on both λ and ΔF we find a different analogy helpful. We pic-
 260 ture a “CO₂ archipelago in a developing, and eventually rising, sea of water-vapor ab-
 261 sorption” (see Figure 4, the poetically inclined might think of these as Planckian out-
 262 croppings in a Simpsonian sea). From this point of view the share of the radiation that
 263 is emitted to space by H₂O in the troposphere, versus that from tropospheric CO₂ or
 264 from the surface, determines the strength of both λ and ΔF . As CO₂ concentrations rise,
 265 or the troposphere deepens, the CO₂ archipelago gains prominence – new islands even
 266 appear with rising CO₂ concentrations, as seen in Seeley and Jeevanjee (2021) – increas-
 267 ing the magnitude of both ΔF and λ . Warming of the atmosphere leads to the devel-
 268 opment of a “sea of absorption”, which progressively reclaims the spectral landscape from
 269 CO₂ and the surface. This reduces ΔF for a given increase in CO₂ and progressively masks
 270 the ability of radiation from the “sea-floor” to escape to space. In our simulations, at
 271 $T_s = 320$ K the “absorption sea-level” is so completely determined by temperature, as
 272 envisaged by Simpson (1928), that the net radiative response to warming in the window
 273 region, $\nu > 767 \text{ cm}^{-1}$, completely vanishes. At this point only the tallest mountains of

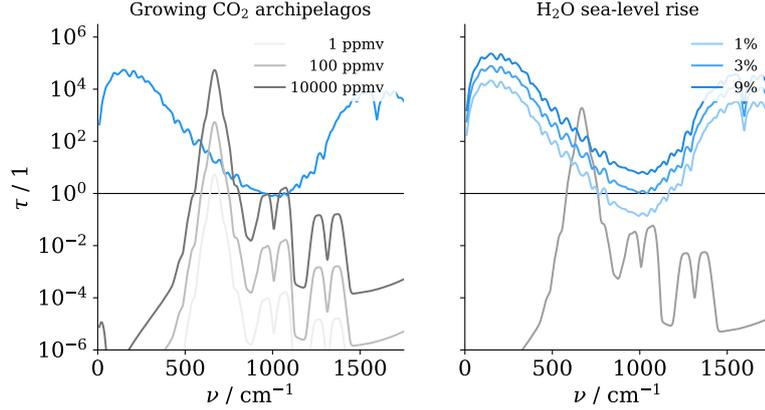


Figure 4. Optical thickness τ as a function of wavenumber ν . The left panel shows how the “CO₂ archipelagos” grow with an increase in atmospheric CO₂ concentration (darker greys). The right panel shows the rising “H₂O absorption sea-level” at higher water vapor volume mixing ratios (darker blues). In addition, the $\tau = 1$ line roughly indicates the location of opaque spectral regions ($\tau > 1$).

274 the “CO₂ archipelago”, whose prominence is pronounced due to a rising tropopause, are
 275 left to balance an increase in forcing.

276 5 Conclusions

277 We perform calculations using the 1D-RCE model konrad at different surface tem-
 278 peratures T_s to analyze the temperature-dependence of the feedback parameter λ for a
 279 fixed CO₂ concentration. A line-by-line treatment of longwave radiant energy transfer
 280 (ARTS) is used to ensure an accurate computation of radiative fluxes and heating rates
 281 over a wide temperature range. By comparison to calculations with the RRTMG radi-
 282 ation scheme, we find that the use of the latter (albeit faster) scheme leads to increas-
 283 ingly erroneous results as surface temperatures increase beyond 300 K – errors in climate
 284 sensitivity are larger than a factor of two at 310 K. This is within the range of temper-
 285 atures sampled by models with very high climate sensitivities subject to quadrupling of
 286 atmospheric CO₂. The erroneous behavior leads to a local maximum (or “bump”) in the
 287 climate sensitivity, similar to what has been found in at least two other modelling stud-
 288 ies (e.g. Meraner et al., 2013; Romps, 2020) using this same, or a similar, treatment of
 289 radiative transfer. The resulting “bump” of the climate sensitivity found in these stud-
 290 ies looks similar to the one predicted by the “CO₂ radiator fin” mechanism by Seeley

291 and Jeevanjee (2021), which arises from the strengthening of the Planck feedback from
 292 more pronounced CO₂ absorption features. In Roms (2020) both effects, the large in-
 293 creases in CO₂, which the climate sensitivity also depends on, and the RRTMG errors
 294 are conflated, and it is unclear which dominates.

295 Using ARTS, λ increases from $-2.1 \text{ W m}^{-2} \text{ K}^{-1}$ to $-1.2 \text{ W m}^{-2} \text{ K}^{-1}$ for T_s between
 296 280 K and 305 K, which can be attributed to a progressive masking of the Planck feed-
 297 back by increased water vapor absorption in the atmospheric window. In our simulations
 298 water vapor completely shuts the atmospheric window at T_s at 320 K, but already by 305 K
 299 this is balanced by a strengthening Planck feedback in the CO₂ absorption band. For
 300 $T_s > 300 \text{ K}$, λ becomes approximately independent of further increases in T_s examined
 301 in this study.

302 For $T_s > 300 \text{ K}$ the radiative forcing ΔF due to CO₂-doubling decreases by about
 303 18 % from a value around 4.5 W m^{-2} to 3.7 W m^{-2} . A spectral analysis of the radiative
 304 forcing reveals that this decrease is caused by increased water-vapor absorption which
 305 masks the radiative forcing at the flanks of the CO₂ absorption band.

306 To help conceptualize these effects we propose the picture of “CO₂ archipelagos
 307 in a sea of water-vapor absorption” to describe the subtle trial of strength between CO₂
 308 and water-vapor absorption. This picture leads to the surprising result that as the at-
 309 mosphere transitions to a moist greenhouse, CO₂ not only becomes less effective as a forcer,
 310 its presence also becomes a prerequisite for maintaining a negative atmospheric feedback.
 311 For these reasons the effect of increasing CO₂ concentrations on Earth’s budget of ter-
 312 restrial radiation alone is incapable of causing a runaway warming – for this to come to
 313 pass, clouds would have to cooperate.

314 **Appendix A**

315 **Model configuration**

316 We are using the 1D radiative-convective equilibrium model konrad (Kluft et al.,
 317 2021, v0.8.1). The boundary conditions are following Kluft et al. (2019) with a CO₂ con-
 318 centration of 348 ppmv. The solar constant is set to 551.58 W m^{-2} at a zenith angle of
 319 42.05° resulting in an insolation of 409.6 W m^{-2} (Wing et al., 2017; Cronin, 2014). The
 320 relative humidity in the troposphere is set to 80 % to ensure a reasonable amount of hu-
 321 midity in the upper troposphere, which is key for the interaction of lapse-rate and water-

322 vapor feedbacks (Minschwaner & Dessler, 2004; Kluft et al., 2019). Above the cold-point
323 tropopause the volume mixing ratio is kept constant.

324 **Radiation scheme**

325 We are using the Rapid Radiative Transfer Model for GCMs (RRTMG Mlawer et
326 al., 1997) through the CliMT Python package. We have checked the radiative fluxes com-
327 puted with CliMT-RRTMG and a stand-alone version and find that they agree within
328 1%. RRTMG is a rapid radiation scheme and uses the distributed- k method for com-
329 putational efficiency. This method requires precalculated lookup tables that are designed
330 to span a wide range of atmospheric states.

331 **Line-by-line treatment of radiation**

332 We are using the Atmospheric Radiative Transfer Simulator (ARTS Eriksson et al.,
333 2011; Buehler et al., 2018). ARTS is a line-by-line radiative transfer model and is used
334 to calculate the longwave radiative fluxes using four emission angles (streams) and based
335 on 32 768 equidistant frequency points between 10 cm^{-1} and 3250 cm^{-1} ($\Delta\nu = 0.1\text{ cm}^{-1}$).
336 Gas absorption is based on the HITRAN database for gas species (Gordon et al., 2017)
337 and additionally the MT_CKD model (Mlawer et al., 2012) for the continuum absorp-
338 tion of water vapor, CO_2 , molecular nitrogen (all Version 2.52), and oxygen (Version 1.00).

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340 Primary data is available on Zenodo through <https://zenodo.org/record/4565196>.

341 konrad v0.8.1 is available at <https://doi.org/10.5281/zenodo.4434837>, and the
342 latest development version can be found at github.com/atmtools/konrad.

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Figure 4.

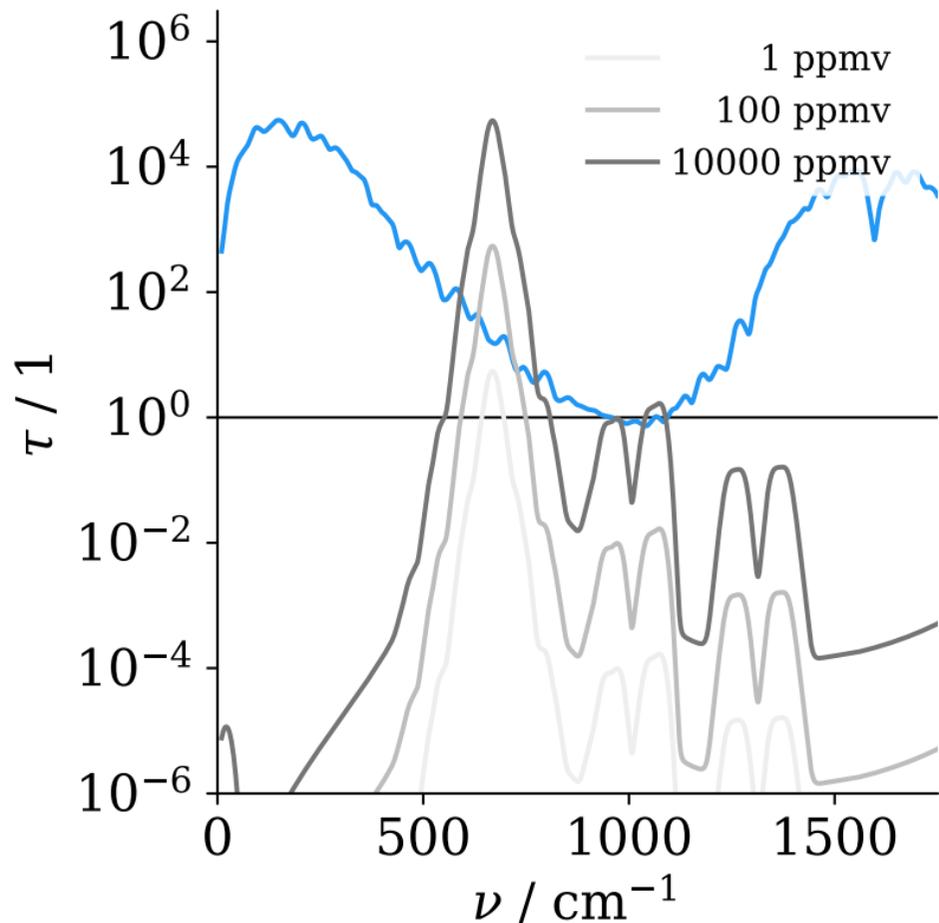
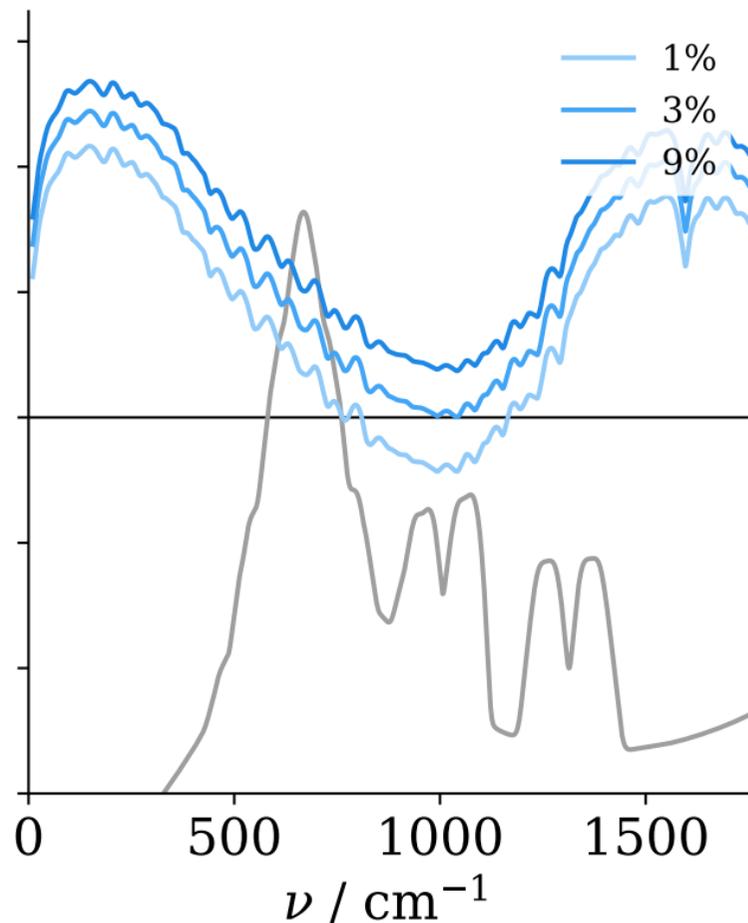
Growing CO₂ archipelagosH₂O sea-level rise

Figure 2.

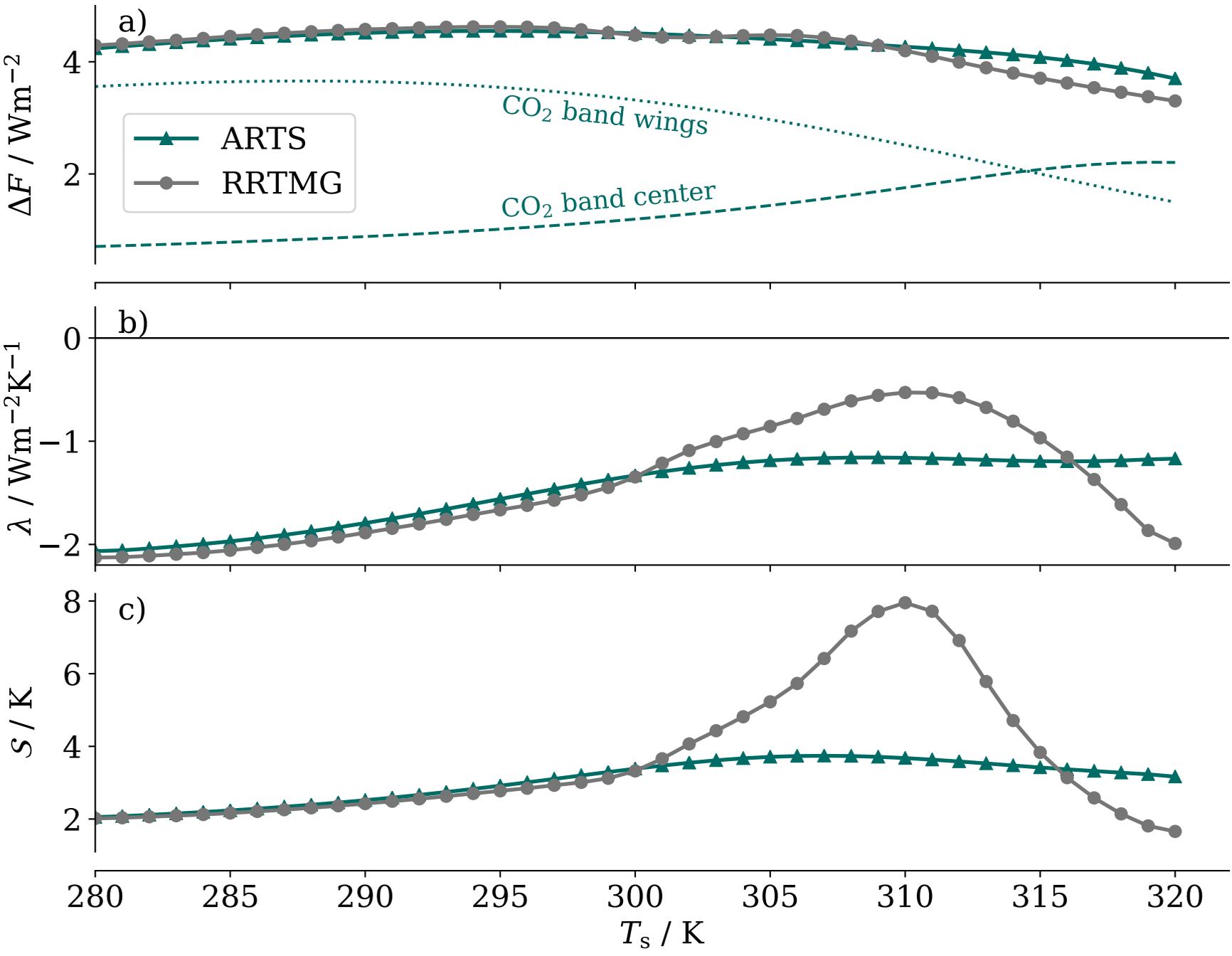


Figure 1.

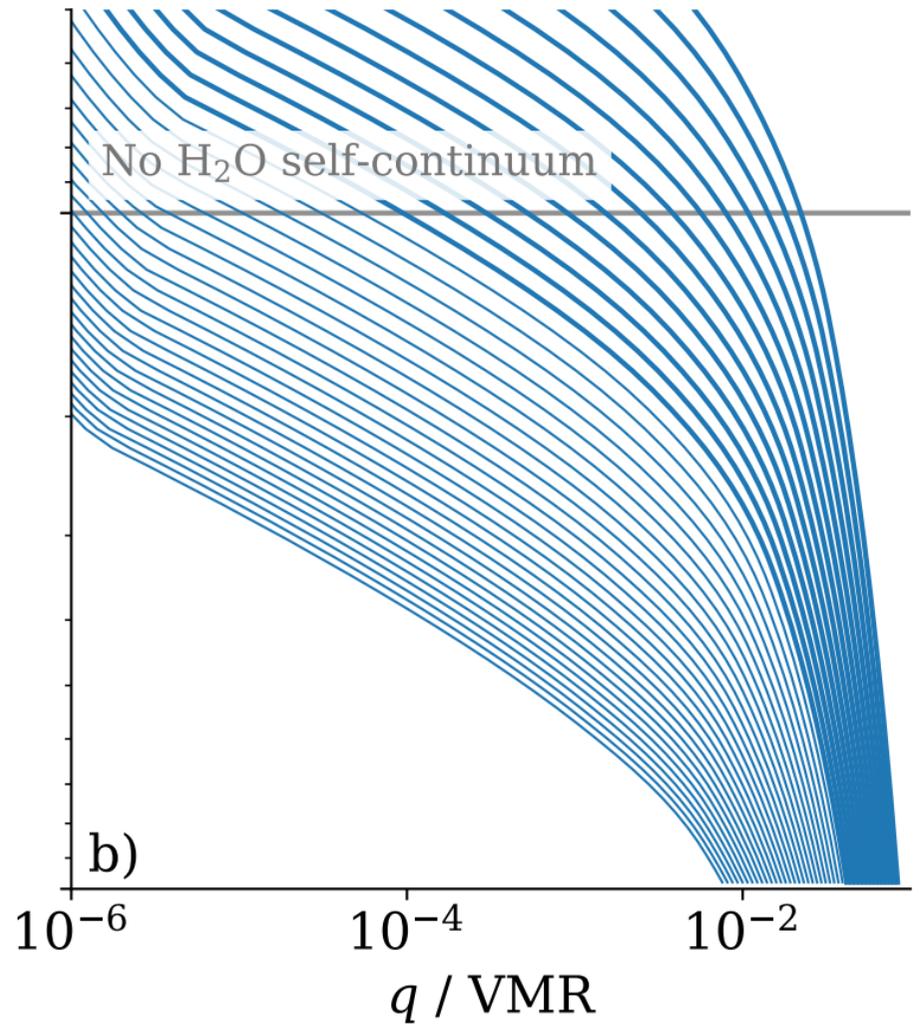
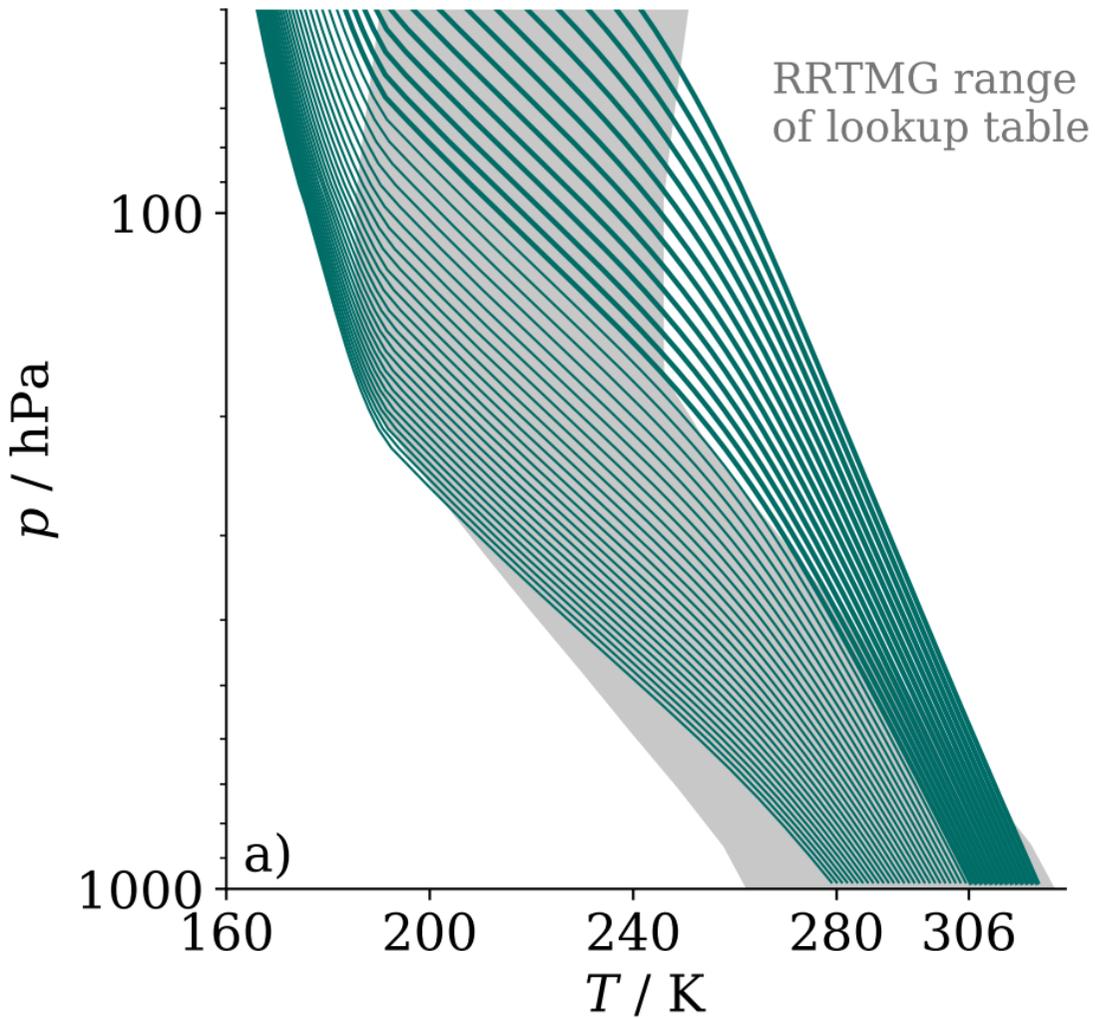


Figure 3.

