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A new set of atmosphere and evolution models for cool T–Y brown dwarfs and giant exoplanets

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ABSTRACT

We present a new set of solar metallicity atmosphere and evolutionary models for very cool brown dwarfs and self-luminous giant exoplanets, which we term ATMO 2020. Atmosphere models are generated with our state-of-the-art 1D radiative-convective equilibrium code ATMO, and are used as surface boundary conditions to calculate the interior structure and evolution of 0.001– $0.075\,M_{\odot}$ objects. Our models include several key improvements to the input physics used in previous models available in the literature. Most notably, the use of a new H–He equation of state including ab initio quantum molecular dynamics calculations has raised the mass by $\sim 1-2\%$ at the stellar–substellar boundary and has altered the cooling tracks around the hydrogen and deuterium burning minimum masses. A second key improvement concerns updated molecular opacities in our atmosphere model ATMO, which now contains significantly more line transitions required to accurately capture the opacity in these hot atmospheres. This leads to warmer atmospheric temperature structures, further changing the cooling curves and predicted emission spectra of substellar objects. We present significant improvement for the treatment of the collisionally broadened potassium resonance doublet, and highlight the importance of these lines in shaping the red-optical and near-infrared spectrum of brown dwarfs. We generate three different grids of model simulations, one using equilibrium chemistry and two using non-equilibrium chemistry due to vertical mixing, all three computed self-consistently with the pressure-temperature structure of the atmosphere. We show the impact of vertical mixing on emission spectra and in colour-magnitude diagrams, highlighting how the $3.5-5.5\,\mu m$ flux window can be used to calibrate vertical mixing in cool T–Y spectral type objects.

Key words. brown dwarfs – planets and satellites: atmospheres – stars: evolution

1. Introduction

The absence or lack of steady hydrogen fusion in the cores of brown dwarfs means that these objects cool over time by radiating away their internal thermal energy. This cooling leads to a degeneracy in mass, age, effective temperature, and luminosity, making the fundamental properties of brown dwarfs, particularly isolated field objects, difficult to determine. The rate at which these objects cool is regulated by the atmosphere, which imprints its complex and changing chemical composition of molecules and condensate species onto the emitted radiation, forming the M-L-T-Y spectral sequence (Kirkpatrick 2005; Helling & Casewell 2014). A reliable model of the atmosphere and its evolution over time therefore lies at the core of our understanding of brown dwarfs and substellar objects. Illustrative of this, the fundamental properties of brown dwarfs are often obtained by fitting synthetic spectra from grids of atmosphere models and then inferring the mass and age of the object using

evolution models (e.g. Saumon et al. 2006, 2007; Burningham et al. 2011; Leggett et al. 2019). Understanding the atmospheres of brown dwarfs has further motivation since the physics, chemistry, and composition is shared with hot Jupiters and directly imaged exoplanets (Burrows et al. 2001), meaning that useful analogies can be drawn between these objects.

Traditionally, the atmospheres of brown dwarfs and giant planets are modelled with 1D codes which solve for the atmospheric temperature structure in radiative-convective flux balance (Marley & Robinson 2015; Fortney 2018). These codes are used to compute grids of models spanning effective temperature and surface gravity containing temperature structures and top of the atmosphere emission spectra for comparison to observations. These atmosphere structures are then coupled as non-grey surface boundary conditions (Chabrier & Baraffe 1997) to interior structure models to compute the cooling and evolution over time. Two of the earliest model sets that follow this framework and that are widely used in the literature include Burrows et al. (1997) and

the AMES-Cond models of Baraffe et al. (2003, hereafter B03). Saumon & Marley (2008, hereafter SM08) presented coupled atmosphere and evolutionary calculations, additionally varying the cloud sedimentation efficiency (Ackerman & Marley 2001) within their atmospheric outer boundary condition, in order to investigate the impact of clouds on brown dwarf evolution. More recently, Fernandes et al. (2019) used existing atmosphere models in the literature as surface boundary conditions to a stellar evolution code to investigate the effects of including additional metals in the interior equation of state (EOS) on the substellar boundary.

Beyond these coupled atmosphere and evolution models numerous improvements and complexities have been added to 1D atmosphere codes in an attempt to reproduce and explain various features of the observed brown dwarf cooling sequence. Cloud models have been developed (Allard et al. 2001; Ackerman & Marley 2001; Helling et al. 2008a) and invoked to explain the reddening L dwarf spectral sequence (Chabrier et al. 2000; Cushing et al. 2008; Stephens et al. 2009; Witte et al. 2011), the sharp change to bluer near-infrared colours at the L-T transition (Allard et al. 2001; Burrows et al. 2006; Marley et al. 2010; Charnay et al. 2018), and the reddening observed in the spectra of late T and Y dwarfs (Morley et al. 2012, 2014). A reduction in the atmospheric temperature gradient has also been explored in 1D models to provide an alternative explanation to the cloudy scenario for this reddening observed along the cooling sequence (Tremblin et al. 2015, 2016, 2017a). This reduction in the temperature gradient has been linked to diabatic convection triggered by the CO/CH₄ transition in brown dwarf atmospheres (Tremblin et al. 2019).

Along with these additional complexities, there has also been significant improvement in the fundamental input physics to 1D atmosphere models. The opacity for important molecular absorbers has improved through more complete hightemperature line lists (Tennyson & Yurchenko 2018), which has altered the temperature structures and synthetic emission spectra in 1D model grids (Saumon et al. 2012; Malik et al. 2019). There has been significant theoretical improvement in the pressure broadened line shapes of the alkali metals Na and K (Allard et al. 2016, 2019), which shape the red-optical and near-infrared spectra of brown dwarfs. Non-equilibrium chemistry due to vertical mixing is a prevalent feature in brown dwarf observations (Noll et al. 1997; Saumon et al. 2000, 2006; Geballe et al. 2009; Leggett et al. 2015, 2017), and theoretical studies have improved our understanding of the impact of such processes in 1D models (Hubeny & Burrows 2007; Zahnle & Marley 2014; Tremblin et al. 2015; Drummond et al. 2016). To further aid in the study of non-equilibrium chemistry processes, complex chemical kinetics networks containing thousands of reactions between important molecules in exoplanet and brown dwarf atmospheres have been developed and refined (Moses et al. 2011; Venot et al. 2012, 2019; Tsai et al. 2017, 2018).

Along with these theoretical improvements, the study of brown dwarfs and giant exoplanets is being driven by everimproving instrumentation that is becoming sensitive to cooler objects. Over the last decade the WISE mission (Wright et al. 2010) has uncovered the coolest spectral type (known as the Y dwarfs; Cushing et al. 2011; Kirkpatrick et al. 2012), including the coldest known brown dwarf at $T_{\rm eff} \sim 250\,\rm K$ just 2 pc from the sun (Luhman 2014). At a few times warmer than Jupiter, these objects provide excellent analogues for Jovian-like worlds outside of our solar system, and are proving challenging for atmosphere models (Morley et al. 2018; Leggett et al. 2019). Ongoing projects are likely to discover more objects in

this temperature range (e.g. Marocco et al. 2019). Accurate and reliable atmosphere and evolutionary models are important for placing mass and age constraints on these newly discovered objects, understanding the rich chemistry and physics taking place in their atmospheres, and determining the low-mass end of the initial mass function (Kirkpatrick et al. 2019).

In this work we present a new set of coupled atmosphere and evolutionary models for brown dwarfs and giant exoplanets. This grid, which we name ATMO 2020, includes numerous improvements to the input physics for modelling substellar objects, and thus supersedes the widely used AMES-Cond grid of B03. We use our 1D atmosphere code ATMO to generate self-consistent models with equilibrium chemistry and non-equilibrium chemistry due to vertical mixing. We include updated line lists for important molecular absorbers and improved line shapes for the collisionally broadened potassium resonance lines. Finally, we couple these atmosphere models to an interior structure model which uses a new H–He EOS from Chabrier et al. (2019) including ab initio quantum molecular dynamics calculations.

The paper is organised as follows. In Sect. 2 we outline the details of the grid and the tools used to generate the models. In Sect. 3 we present the impact of including new potassium resonant line shapes from Allard et al. (2016, hereafter A16) in our 1D atmosphere model ATMO, and compare them to other line shapes available in the literature. Our main results are presented in Sect. 4, where we show how modelling improvements have impacted the predicted cooling tracks, emission spectra, and colours of substellar objects by comparison to other model grids and observational datasets. Finally, we discuss and summarise our work in Sect. 5.

2. Grid set-up and methods

2.1. Model grid

The model set consists of a grid of solar metallicity atmosphere models spanning $T_{\rm eff} = 200-3000\,\rm K$ and $\log(g) = 2.5-5.5\,$ (g in cgs units), with steps of $100\,\rm K$ for $T_{\rm eff} > 600\,\rm K$, $50\,\rm K$ for $T_{\rm eff} < 600\,\rm K$, and 0.5 in $\log(g)$. We note that we extend our grid of models to $T_{\rm eff} = 3000\,\rm K$ in order to follow the evolution of the most massive brown dwarfs from very early stages starting from hot luminous initial models. However, the range of validity of our atmosphere models is $T_{\rm eff} \lesssim 2000\,\rm K$ since we do not include some sources of opacity (e.g. some hydrides and condensates) that form at higher temperatures (see Sects. $2.4\,\rm and$ 5).

We generate three atmosphere grids with different chemistry schemes spanning this parameter range. The first is calculated assuming chemical equilibrium, and the second and third are calculated assuming non-equilibrium chemistry due to vertical mixing with different mixing strengths. Each model in each of the grids is generated with the ATMO code (see Sect. 2.2), and consists of a pressure–temperature (P–T) profile, chemical abundance profiles, and a spectrum of the emergent flux at the top of the atmosphere. These models are publicly available for download^{1,2}.

The P–T profiles from the model atmosphere grid are then used as outer boundary conditions for the interior structure model to follow the evolution of $0.001-0.075\,M_\odot$ objects from $0.001-10\,\mathrm{Gyr}$. We follow the evolution of the object's effective temperature, luminosity, radius, gravity, and absolute magnitudes in a range of photometric filters. Absolute magnitudes are derived by calculating the flux density in a given photometric

http://opendata.erc-atmo.eu

http://perso.ens-lyon.fr/isabelle.baraffe/ATM02020/

filter for each spectrum in the atmosphere grid. The flux density can then be interpolated to the $T_{\rm eff}$ and $\log(g)$ for a given mass and age, and the corresponding radius used to compute the absolute magnitude. The zero point is calculated from a Vega spectrum. The evolutionary tracks for a given mass are also publicly available for download $^{1,\,2}$. We provide more detail on our atmosphere code, chemistry schemes, opacity database, and interior structure model in Sects. 2.2–2.5, respectively.

2.2. One-dimensional atmosphere model: ATMO

ATMO is a 1D–2D atmosphere model developed to study hot Jupiters (Amundsen et al. 2014; Drummond et al. 2016, 2019; Tremblin et al. 2017b; Goyal et al. 2018, 2019b) and brown dwarfs (Tremblin et al. 2015, 2016, 2017a). The model (in 1D) solves for the P–T structure of an atmosphere that is self-consistent with radiative-convective flux balance for a given internal heat flux, and hydrostatic equilibrium for a given surface gravity. This type of model, often termed a radiative-convective equilibrium model, has a long history of being used to study brown dwarf and giant planet atmospheres, and we refer the reader to Marley & Robinson (2015) for a thorough review of these models in this context.

The P-T structure is solved by ATMO on a logarithmic optical depth grid defined in the spectral band between 1.20 and $1.33 \,\mu\text{m}$. We use 100 model levels, with the outer boundary condition in the first model level fixed at a pressure of 10⁻⁵ bar and given an optical depth of $\tau \sim 10^{-4} - 10^{-7}$ depending on $\log(q)$. The inner boundary condition in the last model level is not fixed in pressure and given an optical depth of $\tau = 1000$. A first guess of pressure and temperature is assigned to each model level, and then the model iterates the P-T structure towards radiativeconvective and hydrostatic equilibrium using a Newton-Raphson solver. On each iteration chemical abundances are calculated for the current P-T structure, opacities are obtained from precomputed look-up tables for individual gases, and the radiative and convective fluxes are calculated. The P-T structure is generally considered converged when radiative-convective flux balance and hydrostatic equilibrium is satisfied to an accuracy of $\leq 1 \times 10^{-3}$ in each model level.

ATMO can calculate chemical abundances assuming thermodynamic equilibrium or assuming non-equilibrium chemistry due to vertical mixing in the atmosphere. The chemistry schemes used in this work are discussed in Sect. 2.3. Once the chemical abundances have been computed, the opacities used by ATMO are loaded from pre-computed correlated-*k* tables for individual gases (discussed in Sect. 2.4), and are combined within the code using the random overlap method with resorting and rebinning to get the total mixture opacity (Amundsen et al. 2017). This method ensures the opacities are completely consistent with the pressure, temperature, and abundances on every iteration.

The radiative flux is computed by solving the integral form of the radiative transfer equation in 1D plane-parallel geometry following Bueno & Bendicho (1995). We include isotropic scattering and sample 16 ray directions with a discrete ordinate method using Gauss-Legendre quadrature. The convective flux is computed using mixing length theory (Henyey et al. 1965) using the same method as Gustafsson et al. (2008), with a mixing length of 2 times the local pressure scale height. The adiabatic gradient is computed using EOS tables from Saumon et al. (1995).

2.3. Chemistry schemes

Chemical equilibrium abundances are calculated using a Gibbs energy minimisation scheme based on that of Gordon &

McBride (1994). We use 76 gas phase species (including ionic species) and 92 condensate species with thermodynamic data from McBride et al. (1993, 2002)³. To form these species we include 23 elements: H, He, C, N, O, Na, K, Si, Ar, Ti, V, S, Cl, Mg, Al, Ca, Fe, Cr, Li, Cs, Rb, F, and P. The present models adopt the solar composition of Asplund et al. (2009) with revisions of the elemental abundances of C, N, O, P, S, K, and Fe from the CIFIST project (Caffau et al. 2011). Our equilibrium chemistry scheme has been benchmarked against the GGchem code (Woitke et al. 2018) in Goyal et al. (2019a), and against the Exo-REM and petitCODE 1D atmosphere models in Baudino et al. (2017).

We adopt the rainout approach for the treatment of condensates as described in Goyal et al. (2019a) whereby once a condensate forms, the elements comprising that condensate will be depleted from the current level and all the model levels above (lower pressures). This approach models the settling or sinking of cloud particles in an atmosphere which depletes elemental abundances at lower pressures (Burrows & Sharp 1999). Evidence for this rainout process has been found in the retrieved abundances of alkali metals for late T and Y dwarfs (Line et al. 2017; Zalesky et al. 2019).

To calculate non-equilibrium chemical abundances we have implemented the chemical relaxation scheme of Tsai et al. (2018). Chemical relaxation schemes take the approach of relaxing a species back to its equilibrium abundance, following a perturbation, on a given timescale. The chemical timescale for each species is estimated or parametrised based on a complex chemical kinetics network (Cooper & Showman 2006; Zahnle & Marley 2014; Tsai et al. 2018). Tsai et al. (2018) find the rate-limiting reactions within a chemical network to derive relaxation timescales of H_2O , CO, CO_2 , CH_4 , N_2 , and NH_3 over several P–T regimes in the range 500 to 3000 K, and 0.1 mbar to 1 kbar.

We choose to adopt this chemical relaxation scheme over full chemical kinetics networks for computational efficiency and consistent convergence throughout the grid when solving for a self-consistent P-T profile. The relaxation method is more computationally efficient as it avoids the need to solve the large, stiff system of ordinary differential equations needed when using full chemical kinetics networks. The P-T profile is reconverged on the fly while integrating over time for the non-equilibrium abundances every 50 iterations of the numerical solver. Reconverging the profile more often than every 50 iterations gives negligible differences in the final P-T structure, abundances, and emission spectrum. The chemistry is integrated for a minimum of 1×10^{10} s, and is considered converged and in a steady state when $dn/n < 1 \times 10^{-2}$ and $(dn/n)/dt < 1 \times 10^{-4}$ for all species, where n is the species number density. This self-consistent nonequilibrium chemistry approach is similar to that used in hot Jupiter models presented in Drummond et al. (2016).

Vertical mixing in the atmosphere is parametrised using the eddy diffusion coefficient K_{zz} in cm² s⁻¹, and is assumed to be constant throughout the atmosphere. We scale the eddy diffusion coefficient with surface gravity since the typical dynamical timescale t can be approximated as

$$t \sim \frac{H_P^2}{K_{zz}} \propto \frac{1}{g^2 K_{zz}},\tag{1}$$

where H_P is the atmospheric scale height. Within this approximation, we keep the dynamical timescale t constant by changing

The full list of species is available on http://opendata.ercatmo.eu and http://perso.ens-lyon.fr/isabelle.baraffe/ ATMO2020/

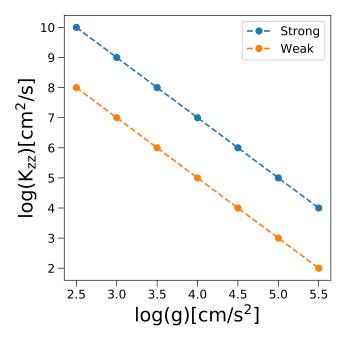


Fig. 1. Vertical mixing relationships with surface gravity (strong and weak; see text) used in the generation of non-equilibrium atmosphere models in this work.

the value of K_{zz} by an order of magnitude for a $\log(g)$ step of 0.5 within the grid. We generate atmosphere model grids with two K_{zz} scaling relationships with surface gravity as shown in Fig. 1; we refer to these relationships as "strong" and "weak" mixing throughout this work.

Our choice of mixing strengths come from approximate values in the literature which have been found to provide reasonable comparisons to observations of late T and Y dwarfs. For example, Leggett et al. (2017) found K_{zz} values in the range 10^4-10^6 cm² s⁻¹ provided reasonable comparison to the [4.5]-M colours of late T and Y dwarfs for model sequences with a constant gravity of $\log(g) = 4.5$ (see their Fig. 7). We have therefore adopted to set $\log(K_{zz}) = 4$ and $\log(K_{zz}) = 6$ in the "weak" and "strong" cases respectively at $\log(g) = 4.5$ and scale K_{zz} with gravity.

We note that K_{zz} has often been estimated by assuming it is the same diffusion coefficient as that derived from mixing length theory of convection, i.e. $D_{\text{mix}} \sim l_{\text{mix}} v_{\text{mlt}}$, with l_{mix} the mixing length and $v_{\rm mlt}$ the convective velocity (Gierasch & Conrath 1985; Ackerman & Marley 2001). This however has to be extrapolated to the convectively stable radiative regions of the atmosphere where a number of complex processes such as gravity waves and convective overshooting (Freytag et al. 1996; Kupka et al. 2018) may drive the mixing. The value of K_{zz} has also been approximated from 3D numerical simulations of hot Jupiters including passive tracer transport (Parmentier et al. 2013; Zhang & Showman 2018b). These approaches to estimating K_{zz} have their limitations and none has provided a quantitative picture that has reached a consensus in the community. In this work we therefore choose to adopt a simpler approximation for K_{zz} to examine the trends of non-equilibrium chemistry in colour-magnitude diagrams (Sect. 4.4), and we leave more sophisticated studies of K_{zz} for future work.

2.4. Opacity database

Our opacity database used by ATMO consists of 22 atomic and molecular species shown in Table 1, and our methodology for

Table 1. Opacity database used by ATMO.

Species	Source
H_2 - H_2 , H_2 - He	Richard et al. (2012)
H^-	John (1988)
H_2O	Barber et al. (2006)
CO_2	Tashkun & Perevalov (2011)
CO	Rothman et al. (2010)
$\mathrm{CH_4}$	Yurchenko & Tennyson (2014)
NH_3	Yurchenko et al. (2011)
Na, K, Li, Rb, Cs, Fe	VALD (Heiter et al. 2015)
TiO	Plez (1998)
VO	McKemmish et al. (2016)
FeH	Wende et al. (2010)
PH_3	Sousa-Silva et al. (2015)
HCN	Barber et al. (2014)
C_2H_2 , H_2S	Rothman et al. (2013)
SO_2	Underwood et al. (2016)

calculating these opacities is presented in detail in Amundsen et al. (2014). The absorption coefficient is calculated on a wavenumber grid spanning 0-50 000 cm⁻¹ and a resolution of 0.001 cm⁻¹, with transitions from the line list sources provided in Table 1. Each line is broadened including both Doppler and pressure broadening with collisions from H2 and He (wherever data is available for each perturber), the dominant species in brown dwarf and hot Jupiter atmospheres. This is done on a pressure and temperature grid, with 40 logarithmically spaced pressure points from $10^{-9}-10^3$ bar, and 20 logarithmically spaced temperature points in the range 70-3000 K. Pressure broadening parameters for H₂ and He are often not provided in the line lists given in Table 1, and are therefore obtained from alternative sources found in Table C.1 of Goyal et al. (2018). The implementation of these pressure broadening parameters and our numerical considerations regarding line wing cutoffs are discussed in Amundsen et al. (2014).

Unlike the molecular line lists used in our database, the VALD line lists for the atomic species contain van der Waals coefficients, which can be used to calculate pressure broadened line widths (Sharp & Burrows 2007). These coefficients are used to calculate widths for all lines except for the Na and K resonance doublets located at ~ 0.59 and $\sim 0.77 \,\mu\text{m}$, respectively, for which a Lorentzian line profile has been shown to be insufficient (Burrows et al. 2000, 2002). The high pressures and temperatures in brown dwarf atmospheres cause these resonance lines to be broadened up to ~4000 cm⁻¹ away from the line centre, and more detailed calculations of these line shapes beyond a Lorentzian profile are required. In previous works with ATMO, we used Na and K line shapes from both Burrows & Volobuyev (2003) and Allard et al. (2007b, hereafter BV03 and A07 respectively). For this work we have updated our K resonant line shapes with those presented in A16. We discuss this in more detail in Sect. 3, and compare the effect that different line shape calculations can have on model brown dwarf atmospheres and synthetic observations.

The radiative transfer equation in ATMO can be solved at the native resolution (0.001 cm⁻¹) of the absorption cross sections (commonly known as the line-by-line approach); however, this is computationally expensive, particularly when iterating for a consistent P–T structure. The more computationally efficient correlated-*k* approximation is therefore used (Lacis & Oinas 1991), which is an approach widely adopted by both the Earth atmosphere and exoplanet communities. The open source UK

Met Office radiative transfer code SOCRATES (Edwards 1996; Edwards & Slingo 1996) is used to generate correlated-*k* opacity tables for each species in our database, and our methodology is described and tested in Amundsen et al. (2014). These tables are computed on the same P–T grid as the full resolution absorption coefficient files, and are provided at 32-, 500-, and 5000-band spectral resolutions. The spacing in the 32-band files is as shown in Table 4 of Amundsen et al. (2014), and these tables are used when iterating for a consistent P–T structure with ATMO, improving computational efficiency while maintaining an accurate heating rate. The 500 and 5000 bands are evenly spaced in wavenumber between 1 and 50 000 cm⁻¹, and the 5000-band tables are used to generate emission spectra shown in this work.

2.5. Interior structure and evolution model

Calculations of interior structure and evolutionary models are based on the Lyon stellar evolution code, and are described in detail in our previous works (Chabrier & Baraffe 1997; Baraffe et al. 1998, 2003). The structure models are based on the coupling between interior profiles and the chemical equilibrium atmospheric structures described previously at an optical depth $\tau = 1000$. We note that this is deeper than our previous models which used $\tau = 100$ to couple the atmosphere to the interior. However, the radial extension of the atmosphere at $\tau = 1000$ is still negligible compared to the total radius of the object, and thus the Stefan-Boltzmann condition ($L = 4\pi\sigma R^2 T_{\text{eff}}^4$) is still satisfied. We use a solar metallicity helium mass fraction Y = 0.275(Asplund et al. 2009) to be consistent with our previous models (Baraffe et al. 1998, 2003; Chabrier et al. 2000). Since we are using a metal-free EOS, the presence of metals with mass fraction Z can be mimicked by an equivalent He mass fraction $Y_{\text{eq}} = Y + Z$ (Chabrier & Baraffe 1997). We use Z = 0.0169 giving

 $Y_{\rm eq} = 0.2919$.

The main change in terms of inner structure input physics concerns the EOS. In this work we use the new EOS for H–He mixtures presented by Chabrier et al. (2019), which includes ab initio quantum molecular dynamics calculations in the regime of pressure dissociation and ionisation. This is a significant improvement over the semi-analytic H–He EOS of Saumon et al. (1995) (SCVH) used in this regime in all our previous models (Baraffe et al. 1998, 2003; Chabrier et al. 2000).

For the sake of comparison, we have also computed a set of evolutionary models with the SCVH EOS to determine the impact of the new EOS. We note that the SCVH EOS is used in the atmosphere models (see Sect. 2.2). There is, however, no difference between the SCVH EOS and the new EOS of Chabrier et al. (2019) in the atmospheric P–T regime, which is close to a perfect gas. There is thus no inconsistency when using the SCVH EOS in the atmosphere models and the new EOS in the inner structure models.

3. Potassium broadening

The alkali metals sodium Na and potassium K play a key role in brown dwarf atmospheres. They are abundant in the gas phase until they condense into KCl and Na₂S (Lodders 1999), and have strong resonance lines at ~ 0.59 and $\sim 0.77 \, \mu m$, respectively, that are present in late L and T dwarfs (Kirkpatrick et al. 1999; Burgasser et al. 2003). The line shapes are determined by the potential field of H₂ perturbing the ground and excited states of the alkali atom, and in brown dwarf atmospheres these resonance lines become broadened out to thousands of angstroms away from the line core, shaping the visible and red-optical

spectra of cool brown dwarfs. As such, Lorentzian line profiles are not sufficient to model the collisional broadening effects on these alkali metals (Allard & Kielkopf 1982; Burrows et al. 2000, 2002; Allard et al. 2019), and more detailed quantum chemical calculations of the interaction potentials of these collisions are required to accurately model Na and K line shapes.

Both BV03 and A07 have presented alkali broadening calculations which can be used in 1D radiative-convective models of brown dwarfs and exoplanets. BV03 calculate the interaction potentials of the ground and excited states of Na and K perturbed by H₂ and He as a function of distance and orientation angle. Using these potentials BV03 computed absorption line profiles using the Franck-Condon model in the quasi-static limit out to thousands of angstroms from the line core. A07 used valence pseudopotentials to compute molecular potentials of Na and K perturbed by H₂ and He, and used the semi-classical unified line shape theory of Allard et al. (1999) to calculate the collisional profiles of the Na- and K-H₂ resonance lines.

Previous works with ATMO have used both the BV03 and A07 broadening treatments as it remains unclear which performs best when reproducing observations. The BV03 profiles used in ATMO are implemented by Baudino et al. (2015). Baudino et al. (2017) benchmarked the BV03 and A07 alkali broadening schemes in a 1D radiative-convective model showing large uncertainties in the predicted transmission spectra of hot Jupiters and the emission spectra of brown dwarfs. When generating the grid of brown dwarf atmosphere models in this work we found similar uncertainties. In particular, the differences in opacity in the far red wing of the K doublet cause substantial differences in the predicted near-infrared spectra where the peak in brown dwarf emission lies (see Fig. 3). This motivated us to implement the new K resonance line profiles presented in A16.

The A16 line profiles follow the same framework as A07, with improvements on the determination of the intermediate-and long-range part of the K–H₂ potential and the inclusion of spin-orbit coupling. The wing profiles of A16 are tabulated for temperatures between 600 and 3000 K as powers of density expansion (Allard et al. 1999). The new line profiles in A16 are valid for $n_{\rm H_2} < 10^{21} \, {\rm cm}^{-3}$, whereas the A07 profiles were valid for $n_{\rm H_2} < 10^{19} \, {\rm cm}^{-3}$, where $n_{\rm H_2}$ is the number density of H₂.

In Fig. 2 we show the absorption cross section of potassium employed in ATMO using broadening schemes from BV03, A07, and A16, at pressures and temperatures typical of the red-optical to near-infrared photosphere of T-type brown dwarfs. The top panel displays the K opacity for P = 1 bar and T = 1500 K. This corresponds to a $n_{\rm H_2}$ < 10^{19} cm⁻³ regime within which both the A07 and A16 profiles are valid. Therefore, the A07 and A16 wing profiles predict a similar strength quasi-molecular K – H₂ line satellite in the blue wing at $\sim 0.7 \,\mu\text{m}$, which is not captured by the BV03 wing profiles. The bottom panel of Fig. 2 shows the K absorption cross section at a higher pressure of 50 bar corresponding to a $10^{19} \,\text{cm}^{-3} < n_{\text{H}_2} < 10^{21} \,\text{cm}^{-3}$ regime within which the A07 tables are no longer valid, while the A16 profiles are. The A07 profiles therefore predict a much weaker line satellite than the A16 profiles. At both 1 and 50 bar, the opacity differs considerably in the red wing at $\sim 1 \,\mu\text{m}$, with the BV03 profiles giving significantly less absorption than the A07 and A16 profiles.

In Fig. 3 we show a synthetic red-optical and near-infrared emission spectrum of a $T_{\rm eff}$ = 800 K, $\log(g)$ = 5.0 T-type brown dwarf calculated with the BV03, A07, and A16 broadening schemes. The red-optical spectra in the left panel shows the difference in the emission around the potassium D1 and D2 resonance doublet. There is a noticeable difference between the

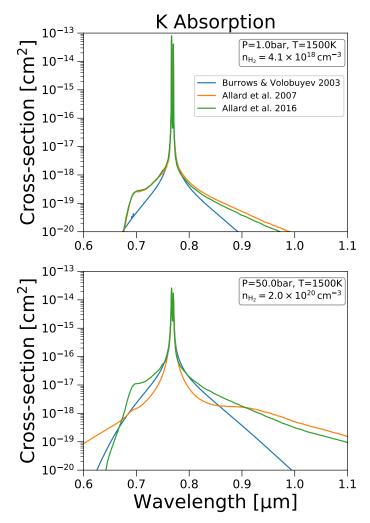


Fig. 2. Absorption cross section of potassium calculated with different broadening treatments for the D1 and D2 resonance doublet, at a pressure of 1 bar and a temperature of 1500 K (*top panel*) and a pressure of 50 bar and a temperature of 1500 K (*bottom panel*).

emission in the blue wing around $\sim 0.7 \, \mu m$ due to the K-H₂ quasi-molecular feature predicted by A07 and A16 compared to BV03. The lower absorption in the red wing in the BV03 case leads to more flux emerging through the Y band at $\sim 1 \, \mu m$ compared to the A07 and A16 cases. The large differences in opacity in the BV03 profiles compared to the A07 and A16 profiles also causes differences in the temperature profile when reconverging the atmospheric structure to find radiative-convective equilibrium. P-T profiles generated including BV03 alkali opacity are several hundred. Kelvin cooler for pressures above 5 bar than profiles generated with A07 and A16 opacity. This leads to the redistribution of flux across the near-infrared seen in. the right panel of Fig. 3. We note that this, flux distribution only occurs if the model is generated, self-consistently with a reconverged P-T structure when switching between opacity sources.

4. Results

This section presents our main results and is organised as follows. In Sect. 4.1 we present our new substellar evolutionary tracks; we highlight the impact of the new EOS, and compare it to other calculations in the literature. In Sect. 4.2 we compare these new evolutionary tracks to dynamical mass measurements

of brown dwarfs. We demonstrate the impact of non-equilibrium chemistry due to vertical mixing on synthetic emission spectra in Sect. 4.3. In Sect. 4.4 we compare our new models to other models and observational datasets in colour-magnitude diagrams. In Sect. 4.5 we make spectral comparisons to other models to highlight improvements in the atmospheric opacities. Finally, in Sect. 4.6 we make initial comparisons of our new models to the observed spectra of cool brown dwarfs across the T–Y transition.

4.1. Evolutionary tracks

In this section we present and compare the new set of atmosphere models and evolutionary tracks to others in the literature in order to highlight model improvements. We choose two families of brown dwarf models that are widely used in the community for comparison, the Lyon group and the Saumon & Marley group.

The Lyon group use the model atmosphere code Phoenix for application to stellar and substellar atmospheres (Allard & Hauschildt 1995; Hauschildt et al. 1999), which have been succesfully used to describe the evolution of low-mass stars (e.g. Baraffe et al. 2015). Both Chabrier et al. (2000) and B03 presented evolutionary calculations for brown dwarfs using grids of Phoenix model atmospheres from Allard et al. (2001), labelled "AMES-Dusty" and "AMES-Cond", respectively. The AMES-Dusty models included dust opacity and are valid for hot (i.e. massive and/or young) brown dwarfs, whereas the AMES-Cond models neglected dust opacity representing the case where all condensates have settled below the photosphere, and are valid for cooler brown dwarfs.

The second set of brown dwarf models we use for comparisons are from the Saumon & Marley group, who applied and developed a 1D radiative-convective code originally designed for solar system atmospheres to brown dwarfs (McKay et al. 1989; Marley et al. 1996, 2002; Burrows et al. 1997). Evolutionary models from this group were presented in SM08, who varied the cloud sedimentation efficiency (see Ackerman & Marley 2001) within their atmospheric outer boundary condition to investigate the impact of clouds on brown dwarf evolution. Here we compare our new model set to the AMES-Cond and SM08 cloud-free models, both of which take the approach used in this work whereby condensate species are included in the chemical equilibrium calculations, but cloud opacity is neglected in the radiative transfer, modelling the scenario where dust grains settle or sediment below the photosphere.

The ATMO and AMES-Cond atmospheric temperature profiles are compared in Fig. 4, for a constant log(g) = 4.0 and $T_{\rm eff}$ between 200 and 2400 K. There are significant differences in the temperatures obtained between the models for a given $T_{\rm eff}$ and $\log(g)$, with the ATMO profiles being typically warmer for $T_{\rm eff}$ < 1200 K and cooler for $T_{\rm eff}$ > 1200 K. There have been numerous model improvements that could contribute to these differences since the AMES-Cond grid was generated. Most notably improved high-temperature line lists including significantly more lines for crucial species such as H₂O, CH₄, and NH₃, have increased the atmospheric opacity leading to warmer temperature profiles for $T_{\rm eff}$ < 1200 K. The ATMO $T_{\rm eff}$ = 200 K model is slightly cooler in the deep atmosphere. This is likely due to the improved treatment of low-temperature equilibrium chemistry and condensation in ATMO, over that used in the Phoenix model at the time of generation of the AMES-Cond grid. For $T_{\rm eff} > 1200 \,\rm K$ the cooler ATMO profiles suggest we may be missing opacity at higher temperatures. We do not include the opacity of some metal oxides and metal hydrides which can be important

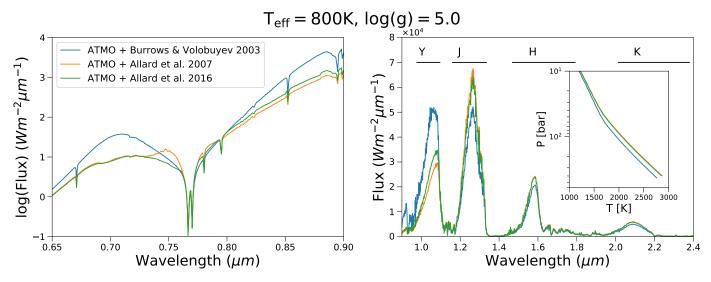


Fig. 3. Comparison of the emission spectra from the top of the atmosphere for $T_{\text{eff}} = 800 \text{ K}$ and $\log(g) = 5.0$ calculated with alkali broadening from BV03; A07; A16. *Right plot*: locations of the Mauna Kea near-infrared filters.

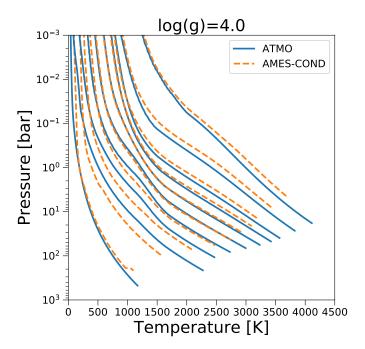


Fig. 4. Self-consistent atmospheric P–T structures from this work (solid blue lines) and from the AMES-Cond models of B03 (dashed orange lines) for $\log(g) = 4.0$ and $T_{\rm eff} = 200, 400, 600, 800, 1000, 1200, 1400, 1600, 2000, and 2400 K (from left to right).$

in shaping the temperature profiles at high $T_{\rm eff}$ (Malik et al. 2019). This is only important for high- $T_{\rm eff}$ objects (i.e. massive and/or young brown dwarfs), and will therefore not affect the evolutionary calculations of cool T–Y objects presented in this work (see further discussion in Sects. 2.1 and 5).

The atmospheric temperature structures from the ATMO model grid are used to couple the non-grey atmosphere to the interior structure, and calculate evolutionary tracks for a range of substellar masses. One of the major improvements of the interior structure model in this work is the use of the EOS of Chabrier et al. (2019), over the older EOS of Saumon et al. (1995) (see Sect. 2.5). Figure 5 shows evolutionary tracks calculated with these different EOSs. There are notable differences for the highest masses, with the new EOS predicting slightly cooler,

less luminous objects at old ages close to the stellar–substellar transition. The new EOS also slightly changes the cooling curve around the deuterium burning minimum mass, which can be seen in the $0.012\,M_\odot$ track.

The right column of Fig. 5 shows evolutionary tracks zoomed in for objects close to the substellar boundary. The largest difference occurs for a 10 Gyr old $0.071\,M_\odot$ object, which is now predicted to be ~180 K cooler in effective temperature and ~0.25 dex less luminous with the new EOS. We note, however, that we do not expect our evolutionary tracks to be accurate at the $0.001\,M_\odot$ level, as other uncertainties in the evolution model such as small changes in the helium mass fraction can cause changes to the cooling curves comparable to those caused by the new EOS. Therefore, distinguishing between the new and the old EOS will be challenging, and for this reason we avoid providing an exact value for the mass at the substellar boundary predicted by our new models.

To illustrate the impact of the new EOS we show the interior temperature and density profiles of a 0.075 M_{\odot} , 10 Gyr object in Fig. 6. The new EOS of Chabrier et al. (2019) gives an object up to ~5% cooler and ~8% denser in the core. This therefore raises the theoretical stellar–substellar boundary by 1–2% in mass, as the interior is now cooler and denser, thus more degenerate ($\psi \propto T/\rho^{2/3}$, where ψ is the degeneracy parameter, Chabrier & Baraffe 1997) for a given mass and age. This results in a change in the cooling curves at masses near the stellar–substellar boundary, with objects cooling to lower $T_{\rm eff}$.

The evolutionary tracks from this work and from the AMES-Cond and SM08 calculations are compared in Fig. 7, which shows the evolution of the effective temperature and luminosity for masses between 0.001 and $0.075\,M_\odot$. As previously discussed, the new EOS used in this work raises the hydrogen and deuterium minimum burning mass, causing changes in the cooling curves at high masses and around $0.012\,M_\odot$, respectively. Furthermore, the shape of the evolutionary tracks of the lowest masses have changed due to the differences in the atmospheric temperature structures used as the outer boundary condition. The warmer ATMO temperature structures for $T_{\rm eff} < 1200\,{\rm K}$ (Fig. 4) lead to a slightly cooler, less luminous $0.001\,M_\odot$ object for ages $< 0.1\,{\rm Gyr}$, and a warmer, brighter object for ages $> 0.1\,{\rm Gyr}$. We find qualitatively similar differences when

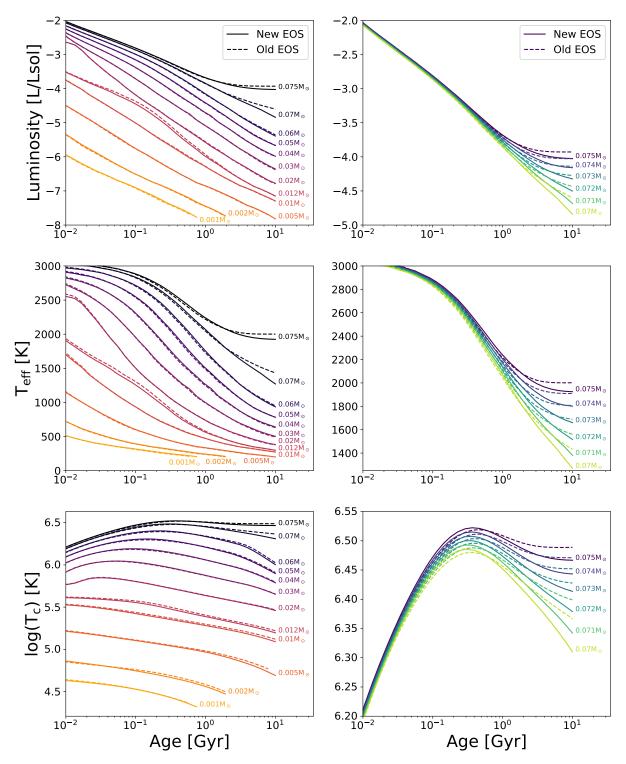


Fig. 5. Luminosity, effective, and central temperature (*from top to bottom*) as a function of age calculated with the new EOS from Chabrier et al. (2019) (solid lines) and the older EOS of Saumon et al. (1995) (dashed lines). These models are calculated with ATMO surface boundary conditions, as described in the text. Each column displays a different selection of masses, indicated by colour-coded annotations placed next to the curves calculated with the new EOS.

comparing ATMO to the AMES-Cond and to the SM08 tracks, respectively.

4.2. Comparison with dynamical masses

Dynamical mass measurements of brown dwarfs from astrometric monitoring programs of binary systems provide useful tests

for evolutionary models (e.g. Dupuy & Liu 2017). Recently, Brandt et al. (2019) presented a dynamical mass measurement of the first imaged brown dwarf Gl 229 B of $70 \pm 5 \, M_{\rm Jup}$. This measurement joins a growing list of massive T dwarfs that are challenging evolutionary models (e.g. Bowler et al. 2018; Dieterich et al. 2018; Dupuy et al. 2019). We note, however, that the dynamical mass measurement of Gl 229 B should

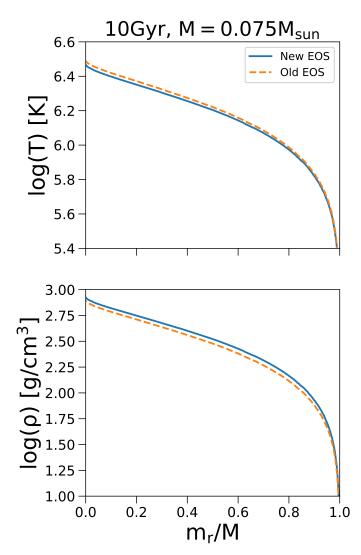


Fig. 6. Interior temperature (*top*) and density (*bottom*) as a function of normalised radial mass profile of the simulated object. Solid blue lines are generated using the new EOS of Chabrier et al. (2019), and the dashed orange lines are generated using the older EOS of Saumon et al. (1995).

be considered with caution until confirmed unambiguously (R. Oppenheimer, priv. comm.; Brandt et al. 2019).

We show in Fig. 8 the luminosity as a function of mass for ultracool dwarfs with dynamical mass measurements including Gl 229 B. In this figure we show isochrones from this work calculated with the new and old EOSs, and isochrones from B03. The new EOS can be seen predicting cooler, less luminous objects in this figure for old high-mass objects. For a 70 $M_{\rm Jup}$ object at an age of 10 Gyr, the ATMO tracks calculated with the new EOS are ~ 0.1 dex less luminous than the AMES-Cond tracks of B03, and ~ 0.4 dex less luminous than the hybrid cloud tracks of SM08.

As discussed by Brandt et al. (2019), the evolutionary models of B03 and SM08 are only compatible with a mass of $70 \pm 5 \, M_{\rm Jup}$ for Gl 229 B if the system is old (7–10 Gyr), in some tension with the 2–6 Gyr age estimate of the host star from kinematics and stellar activity. The ATMO tracks calculated with the new EOS may help relieve some of the tension surrounding the age of the system given that high-mass objects are predicted to be cooler and less luminous at a given age. We note, however, that the difference between the old and new EOS is not observationally

significant given the uncertainty on the mass measurement of GI 229 B shown in Fig. 8.

4.3. Equilibrium and non-equilibrium chemistry models

Non-equilibrium processes primarily affect the carbon and nitrogen chemistry of the atmosphere of a cool T–Y-type object (Zahnle & Marley 2014). Mixing processes can be responsible for bringing CO and N_2 into the upper atmosphere where CH₄ and NH₃ should be the dominant carbon- and nitrogenbearing species according to thermodynamic equilibrium. The chemical timescale to convert CO \rightarrow CH₄ and $N_2 \rightarrow$ NH₃ is typically long compared to mixing timescales, meaning excess CO is predicted in T-type objects (Fegley & Lodders 1996) and depleted ammonia in cooler late T–Y-type objects (Zahnle & Marley 2014; Tremblin et al. 2015). Indeed, observational studies have revealed an excess of CO in both the atmosphere of Jupiter (Bézard et al. 2002) and T dwarfs (Noll et al. 1997; Geballe et al. 2009), and depleted NH₃ in late T and Y dwarfs (Saumon et al. 2000, 2006; Leggett et al. 2015; Tremblin et al. 2015).

Figures 9 and 10 show chemical abundance profiles and synthetic emission spectra, respectively, calculated assuming chemical equilibrium (CEQ) and consistent non-equilibrium chemistry (CNEQ) due to vertical mixing for a sample of effective temperatures and surface gravities. We note that, as discussed in Sect. 2.3, we scale the eddy diffusion coefficient K_{zz} with surface gravity such that vertical mixing is stronger in lower gravity objects. As such the differences in the spectra presented in Fig. 10 are larger for lower gravity models. Furthermore, in these figures we show non-equilibrium models calculated with the strong K_{zz} mixing relationship (Fig. 1), to maximise the differences between the equilibrium and non-equilibrium spectra.

The abundance of CH₄ is quenched by approximately an order of magnitude in the upper atmosphere for the $T_{\rm eff}$ = 800 K, log(g) = 3.5 model in Fig. 9. The depleted CH₄ abundance lowers the opacity in the absorption bands at ~ 1.6 and $\sim 3.15 \,\mu m$ giving brighter H and L' bands in the non-equilibrium spectrum of this model in Fig. 10. The K-band flux is lower in the non-equilibrium spectrum due to the P-T profile being ~150 K cooler at 1 bar than the equilibrium model. This causes the model levels in which the K-band flux is generated to be shifted to slightly higher pressures where the H₂-H₂ collisionally induced absorption is stronger. The weaker mixing in the high-gravity $T_{\rm eff} = 800 \, \text{K}$, $\log(q) = 5.0 \, \text{model means that CH}_4$ is not depleted as strongly (Fig. 9). This smaller change in CH₄ abundance combined with the near-infrared photosphere lying deeper in the atmosphere for higher gravity means that there is no change in the spectrum in the H, K, and L' bands, as there was in the lower gravity $T_{\text{eff}} = 800 \text{ K}$, $\log(g) = 3.5 \text{ model}$.

The abundances of CO and CO₂ are increased by many orders of magnitude in the upper atmosphere under non-equilibrium chemistry due to vertical mixing in all models shown in Fig. 9. Despite CO₂ being several orders of magnitude less abundant than CO, both CO and CO₂ have strong absorption features at ~4.3 and ~4.18 μ m, respectively, and their increased abundances lower the flux at these wavelengths in the *W*2 and *M'* bands in the non-equilibrium spectra shown in Fig. 10. The abundance of NH₃ is quenched in the models shown in Fig. 9 under non-equilibrium chemistry, and can be seen having an effect on the predicted spectrum of the cooler $T_{\rm eff}$ = 400 K models shown in Fig. 10. The depleted NH₃ abundance lowers the opacity in the NH₃ absorption bands at ~1.5 and ~2.85 μ m giving a brighter *H* band and more flux at ~3 μ m in the non-equilibrium spectrum.

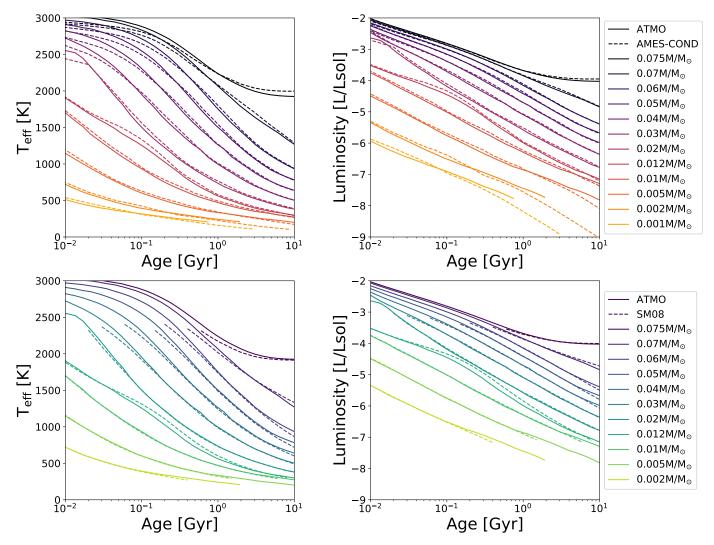


Fig. 7. Evolution of the effective temperature and luminosity for a range of substellar masses from this work (solid lines), from the AMES-Cond models of B03 (dashed lines, *top*), and the SM08 models (dashed lines, *bottom*).

4.4. Colour-magnitude diagrams

In this section we compare the new model set presented in this work, along with the models of B03 and SM08, to observational datasets in colour-magnitude diagrams. In each of the colour-magnitude diagrams presented in Figs. 11–14, the left panel shows isochrones of photometry derived from chemical equilibrium atmosphere models from each of these three works, and the right panel shows isochrones of the equilibrium and non-equilibrium models of this work to illustrate the impact of vertical mixing on the predicted colours of cool brown dwarfs. In Figs. 11 and 12 we present near-infrared colour-magnitude diagrams including photometry from the database of ultracool parallaxes (Dupuy & Liu 2012; Dupuy & Kraus 2013). We exclude from the dataset the known and suspected binaries, young low-gravity objects, and low-metallicity objects.

Figure 11 shows the J-H colours as a function of absolute J magnitude. The data show the M and L dwarf population for J < 14, which gets progressively redder for cooler objects, and the sharp change to bluer colours for the methane dominated T dwarfs at $J \sim 14.5$ known as the L-T transition. The cool T and Y dwarf objects for which the models presented in this work are most applicable, lie below $J \sim 15$, and their J-H colours are best reproduced by the AMES-Cond models, with both the ATMO

and SM08 isochrones predicting colours that are too blue compared to the data. However, this is caused by the outdated physics used within the AMES-Cond models, which lack CH_4 and NH_3 opacity in the H band due to the incomplete line lists used at the time. The brighter AMES-Cond H band therefore gives redder J-H colours that coincidentally more closely match the data compared to the ATMO and SM08 tracks, which use more complete CH_4 and NH_3 line lists. These more complete line lists have added opacity to the H band since the generation of the AMES-Cond models.

As discussed in Sect. 4.3 and shown in Fig. 10, non-equilibrium chemistry due to vertical mixing can quench CH_4 and NH_3 , lowering the opacity and increasing the flux through the H band compared to models calculated in chemical equilibrium. As shown in the right panel of Fig. 11, vertical mixing reddens the predicted J-H colours compared to chemical equilibrium tracks, moving the isochrones towards the observed colours of T dwarfs. The difference between the weak and strong vertical mixing tracks is small. This is due to the abundance of methane only varying by a small factor at the quench levels corresponding to the differing K_{zz} values, hence giving similar H band magnitudes. Despite the non-equilibrium chemistry models improving the J-H colours through H-band brightening, additional physics not included in this work such as reductions

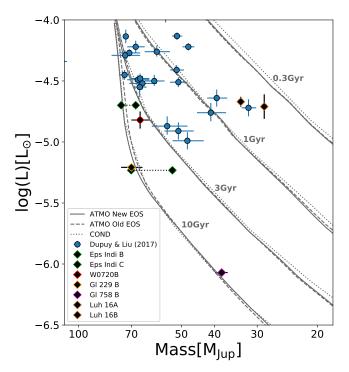


Fig. 8. Luminosity as a function of mass for ultracool dwarfs that have dynamical mass measurements. ATMO isochrones from this work calculated using the new EOS from Chabrier et al. (2019) and the older EOS of Saumon et al. (1995) are plotted as solid and dashed grey lines respectively. Isochrones from the AMES-Cond models of B03 are plotted as dotted grey lines. Most mass measurements are from Dupuy & Liu (2017) and are plotted as blue circles, with other literature measurements plotted as black diamonds (Bowler et al. 2018; Lazorenko & Sahlmann 2018; Dupuy et al. 2019; Brandt et al. 2019). Both mass measurements of ϵ Indi BC from Cardoso (2012) and Dieterich et al. (2018) are plotted, and we refer the reader to Dupuy et al. (2019) for a discussion on these conflicting mass measurements. We indicate key objects with coloured outlines.

in the temperature gradient due to thermochemical instabilities (Tremblin et al. 2015, 2016) and/or cloud opacity (Morley et al. 2012) can reduce the flux in the J band and better reproduce the red J-H colours of late T dwarfs. For cooler objects, the chemical equilibrium tracks begin to redden and reconverge with the observed colours of the Y dwarfs, which lie below $J \sim 21$ in Fig. 11.

Figure 12 shows the H-K colours as a function of absolute H-band magnitude. Similarly to Fig. 11, the reddening M-L sequence along with the sharp L-T transition to bluer colours is shown by the data. The cool T-Y-type objects lying below $H \sim 15$ have H-K colours best reproduced by the ATMO isochrones, with the AMES-Cond and SM08 tracks predicting colours that are too blue compared to the data. In addition to the differing CH₄ and NH₃ opacity and line lists used in the models causing differences in the H band, the updated H_2-H_2 collisionally induced absorption used in the ATMO models is responsible for altering the K-band magnitude and improving the comparison with the observations in this diagram (see also Saumon et al. 2012). Unlike the J-H colours in Fig. 11, including non-equilibrium chemistry due to vertical mixing moves the isochrones away from the observed H-K colours. This is due to the quenching of CH₄ and NH₃ brightening the H band, as shown in Fig. 10. Again, the difference between the weak and strong vertical mixing tracks is small, due to the reasons mentioned above. Similarly to the J-H colours, the H-K colours could be improved by temperature gradient reductions and/or cloud opacity, neither of which is included in this work.

As noted by Leggett et al. (2019), cool T-Y-type brown dwarfs emit a large percentage of their total energy through the $3.5-5.5 \mu m$ flux window, at longer wavelengths probed by the J, H, and K filters considered in Figs. 11 and 12. The WISE W1 and W2 filters probe this wavelength region and can provide useful photometry by which to characterise cool brown dwarfs. We show the H versus H–W2 colour-magnitude diagram in Fig. 13, which contains data points from Kirkpatrick et al. (2019). Kirkpatrick et al. (2019) presented new measurements of trigonometric parallaxes of late T and Y dwarfs with the Spitzer space telescope, and combined these measurements with others published in the literature to complete a sample of ≥T6 dwarfs within 20 pc. Within this sample Kirkpatrick et al. (2019) found a tight correlation of the H-W2 colour with absolute H-band magnitude (see their Fig. 8), which provides a useful metric for benchmarking atmosphere models.

The H-W2 colours in Fig. 13 are nicely reproduced by the ATMO isochrones in comparison to the AMES-Cond tracks, which become too blue for objects fainter than ~19 magnitude due to missing opacity in the H band from the incomplete CH₄ and NH₃ line lists used. The ATMO chemical equilibrium tracks, even though a great improvement over the AMES-Cond models, are slightly too red compared to the data. Including nonequilibrium chemistry due to vertical mixing reduces the W2 band magnitude due to increased CO and CO₂ absorption and increases the H-band magnitude due to the quenching of CH₄ and NH₃ (see Fig. 10), resulting in bluer H-W2 colours. The weak vertical mixing tracks provide the best comparison to the data; the strong mixing tracks are slightly too blue compared to the data. This indicates that the H-W2 colours of brown dwarfs can be used to calibrate vertical mixing and constrain the values of K_{zz} that should be used in atmosphere models.

A long-standing issue in the understanding of cool brown dwarfs is known as the 4 micron problem, whereby the $\lambda \sim 4 \,\mu m$ model fluxes are too low compared to observations of cool brown dwarfs (Leggett et al. 2012, 2013, 2015, 2017). This problem has most recently been demonstrated and discussed by Leggett et al. (2019), who presented new L' photometry of a sample of late T and Y dwarfs showing that the $\lambda \sim 4\mu m$ discrepency occurs in objects cooler than $T_{\rm eff} \sim 700$ K and increases towards lower $T_{\rm eff}$. We show in Fig. 14 the H-L' colours as a function of absolute L'band magnitude with photometric data points from Leggett et al. (2019). All the chemical equilibrium models in the left panel of Fig. 14 underpredict the L' magnitude for H-L' > 3, corresponding to objects with $T_{\rm eff}$ < 600 K using the ATMO 1 Gyr isochrone. Including non-equilibrium chemistry due to vertical mixing can increase the $\lambda \sim 4 \,\mu \text{m}$ flux as CH₄ is quenched in the atmosphere, lowering the opacity at this wavelength (see Fig. 10). However, as the H band also brightens when including vertical mixing the H-L' colours become bluer, moving the tracks away from the observed population of late T and Y dwarfs compared to the chemical equilibrium tracks. As noted by Leggett et al. (2019) and Morley et al. (2018), the discrepancy between the models and the observed $\lambda \sim 4 \,\mu \text{m}$ flux is likely due to processes happening in these atmospheres that are not currently captured by 1D radiative-convective models, such as thermochemical instabilities, cloud clearing, or breaking gravity waves.

4.5. Spectral comparisons with other models

The AMES-Cond grid of the Lyon group was labelled as such due to the NASA-AMES line lists used to calculate the opacity

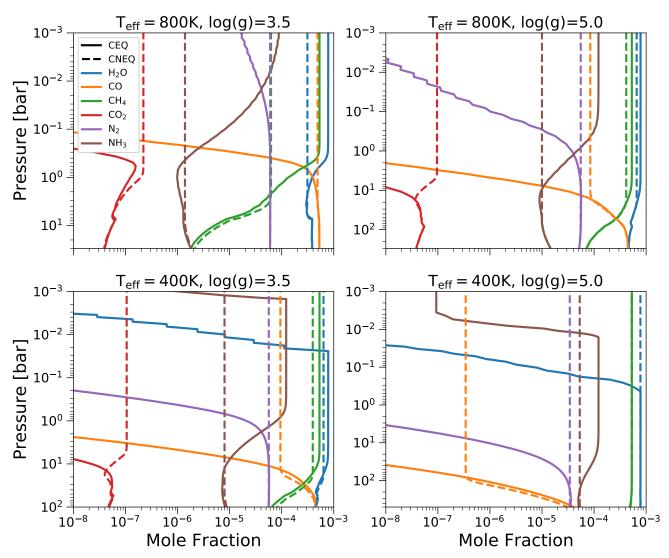


Fig. 9. Chemical abundance profiles of H_2O , CO, CH_4 , CO_2 , N_2 , and NH_3 of self-consistent ATMO models generated under the assumption of chemical equilibrium (solid lines) and non-equilibrium chemistry due to vertical mixing (dashed lines). Non-equilibrium models are calculated with the strong K_{zz} mixing relationship with surface gravity, as shown in Fig. 1. *Rows*: models with different effective temperatures, *columns*: models with different surface gravities, as indicated in the plot titles. In the top left $T_{eff} = 800 \text{ K}$, $\log(g) = 3.5 \text{ plot}$, the non-equilibrium abundance of CH_4 lies below the N_2 abundance.

of $\rm H_2O$ and TiO. Since the calculation of these models there have been significant improvements in high-temperature line lists for these species, in particular the BT2 $\rm H_2O$ line list from Barber et al. (2006). A new BT-Cond grid of Phoenix model atmospheres with updated opacities was presented by Allard et al. (2012), which spans $T_{\rm eff} = 800-3000\,\rm K$. Figure 15 shows comparisons of emission spectra from ATMO and BT-Cond for a selection of effective temperatures. Differences in the emission spectra can be seen in the H and K bands due to the updated CH₄ line list and improved $\rm H_2-H_2$ collisionally induced absorption used by ATMO. Furthermore, for $T_{\rm eff} = 800\,\rm K$ differences can be seen in the Y band at $\lambda \sim 1\,\mu\rm m$, likely due to different potassium broadening schemes (see Sect. 3).

Saumon et al. (2012, hereafter S12) also presented updated atmosphere models from the Saumon & Marley group including improved NH₃ and H₂ opacities, for $T_{\rm eff} = 300-1500\,\rm K$. We compare the infrared emission spectra predicted by ATMO against spectra from the S12 grid in Fig. 16, finding good overall agreement between the models, particularly in the $3.5-5.5\,\mu\rm m$ flux window. Similarly to the comparisons with the BT-Cond grid,

differences lie in the CH₄ absorption band at $\lambda \sim 1.6\,\mu m$ due to the updated line list used by ATMO. Further differences arise at lower $T_{\rm eff}$ within an NH₃ absorption band at $\lambda \sim 1\,\mu m$. Both the ATMO and S12 models use the same ExoMol line list from Yurchenko et al. (2011), meaning differences are likely due to differing NH₃ abundances and condensation treatments in the models. We note that further spectral differences between ATMO and the BT-Cond and S12 models in Figs. 15 and 16 are likely due to discrepancies in the P–T profiles brought about by differing opacity sources impacting the temperature structure.

4.6. Spectral comparisons with observations

In Figs. 17 and 18 we show comparisons of our models to spectra and photometry of cool T–Y-type brown dwarfs. Our methodology here is to compare models to the data by eye, guided by values of $T_{\rm eff}$, $\log(g)$, and R obtained from other studies in the literature and which are consistent with our new evolutionary tracks. Our by-eye comparison serves to illustrate model improvements and current shortcomings in reproducing cool

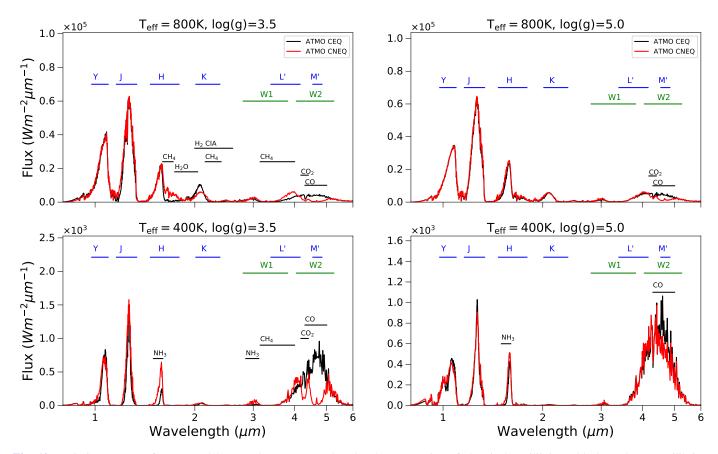


Fig. 10. Emission spectra of ATMO model atmospheres generated under the assumption of chemical equilibrium (black) and non-equilibrium chemistry due to vertical mixing (red). Non-equilibrium models are calculated with the strong K_{zz} mixing relationship with surface gravity, as shown in Fig. 1. *Rows*: models with different effective temperatures, *columns*: models with different surface gravities, as indicated in the plot titles. Overplotted for clarity are the approximate locations of molecular absorption features causing differences between the equilibrium and non-equilibrium spectra. Also indicated in the plots are the locations of the Mauna Kea near-infrared photometric filters (blue bars), and the WISE infrared filters (green bars).

brown dwarf spectra, and we leave more thorough grid fitting analyses to future work.

Gliese 570 D is a late T dwarf companion to a ternary star system ~5.8 pc parsecs away from the sun (Burgasser et al. 2000; van Leeuwen 2007). It has a T7.5 spectral type and is one of the most thoroughly studied T dwarfs to date. Age indicators from the host star indicate an age in the range 1–5 Gyr (Geballe et al. 2001; Liu et al. 2007). Gliese 570 D has been the target of a number of grid fitting studies, which have estimated $T_{\rm eff} = 800-820 \, {\rm K}$, $\log(g) = 5.00-5.27$ and $L = 2.88-2.98 \, L_{\odot}$ (Geballe et al. 2001, 2009; Saumon et al. 2006, 2012). This object has also been used as a benchmark for brown dwarf retrieval studies, which obtain a slightly cooler $T_{\rm eff} = 715 \, {\rm K}$ and a surface gravity $\log(g) = 4.8$ (Line et al. 2015, 2017). Red-optical and near-infrared spectra are from Burgasser et al. (2003, 2004).

We compare $T_{\rm eff} = 800$ K, $\log(g) = 5.0$ chemical equilibrium models calculated with different K resonance line broadening schemes to the red-optical and near-infrared spectra of Gliese 570 D (Burgasser et al. 2003, 2004) in Fig. 17. We find a radius of $R/R_{\odot} = 0.082$ provides the best match to the observed spectrum for this $T_{\rm eff}$ and $\log(g)$. Using our new evolutionary tracks, these parameters indicate an age of 5 Gyr and a mass of $46\,M_{\rm Jup}$ for Gliese 570 D, in agreement with previous works (Saumon et al. 2006). We note that non-equilibrium chemistry models do not impact the near-infrared spectrum within this wavelength range since the $K_{\rm zz}$ value is low for this high gravity.

The models with A16 K resonance line broadening provide the best match to the data. There is an excellent agreement in the Y band where the redwing of the K resonance doublet influences the spectrum. In models with A07 K broadening the opacity in the redwing is too strong giving too little flux in the Y band, whereas in models with BV03 broadening too much flux emerges in the Y band due to the lower opacity in the K redwing. A further improvement in the models can be seen in the H band, where the improved methane line list provides a much more satisfactory comparison to the data than models with a less complete line list (Saumon et al. 2012). The K band is nicely reproduced due to the collisionally induced absorption from Richard et al. (2012), as previously shown in Saumon et al. (2012). The flux in the J band is overpredicted by the model with A16 K broadening. We speculate that a more in-depth fitting study investigating the effects of metallicity and/or thermo-chemical instabilities may help further improve the fit in the J band.

In Fig. 18 we show a comparison of our models calculated with equilibrium and non-equilibrium chemistry to spectra and photometry of objects spanning the T–Y transition. We compare them to the T9 spectral standard UGPS 0722 (Lucas et al. 2010; Leggett et al. 2012), a well-studied cool dwarf that has been estimated to have $T_{\rm eff} = 505 \pm 10 \, {\rm K}$, a mass of 3–11 $M_{\rm Jup}$, and an age range between 60 Myr and 1 Gyr using the SM08 models (Leggett et al. 2012). We compare $T_{\rm eff} = 500 \, {\rm K}$, $\log(g) = 4.0 \, {\rm Chemical}$ equilibrium and non-equilibrium models to this object, finding that these models overpredict the flux in the Y and

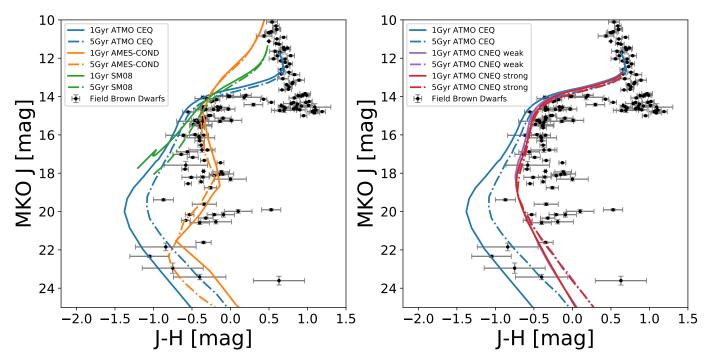


Fig. 11. *J*–*H* colour as a function of absolute *J*-band magnitude. *Left*: isochrones from chemical equilibrium models of this work (blue), B03 (orange), and SM08 (green). *Right*: isochrones from chemical equilibrium models (blue) and non-equilibrium chemistry models with weak (purple) and strong (red) vertical mixing from this work. In both panels the photometry of field brown dwarfs are plotted as black circles, with the data taken from the database of ultracool parallaxes. The data is filtered to only include field brown dwarfs from Dupuy & Liu (2012) and Dupuy & Kraus (2013).

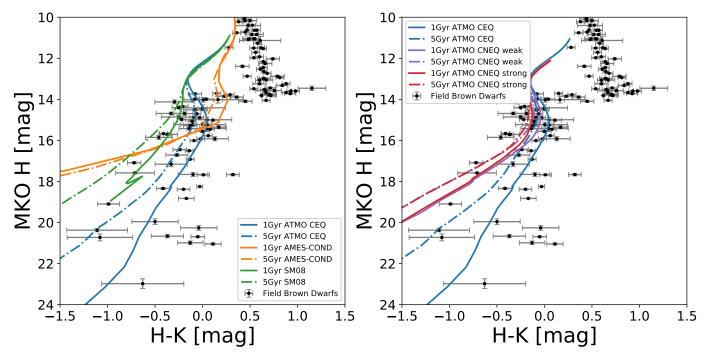


Fig. 12. Same as Fig. 11, but for the H-K colour as a function of absolute H-band magnitude. Isochrones from chemical equilibrium models are compared in the *left panel*, and the effect of non-equilibrium chemistry on isochrones is shown in the *right panel*.

J bands at $\sim 1.0 \, \mu \mathrm{m}$ and $\sim 1.2 \, \mu \mathrm{m}$, respectively. This has been noted by other authors (e.g. Leggett et al. 2012), with sulfide clouds (Morley et al. 2012) or a reduced temperature gradient (Tremblin et al. 2015) invoked to redden the spectrum at these short near-infrared wavelengths. At longer wavelengths, the shape of the K band at $\sim 2.1 \, \mu \mathrm{m}$ appears to be better reproduced by the model including non-equilibrium chemistry. The

Spitzer IRAC channel 2 and WISE W2 photometric points at $\sim 4.5 \, \mu \text{m}$ and $\sim 4.6 \, \mu \text{m}$, respectively, are lower than that predicted with the chemical equilibrium model, implying the presence of enhanced CO absorption brought about through vertical mixing in the atmosphere (see Fig. 10 and Sect. 4.3). Both the strong and weak mixing non-equilibrium models overpredict the CO absorption in the IRAC ch2 and W2 bands, implying that the

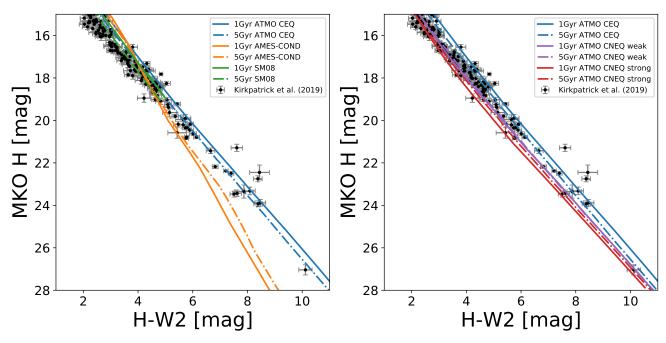


Fig. 13. *H*–*W*2 colour as a function of absolute *H*-band magnitude, with photometry of late T and Y dwarfs within 20 pc from Kirkpatrick et al. (2019) plotted as black circles. Overplotted in the *left panel* are isochrones of chemical equilibrium models from this work (in blue), from B03 (in orange), and from SM08 (in green). Overplotted in the *right panel* are isochrones of chemical equilibrium models (in blue), and non-equilibrium chemistry models with weak and strong vertical mixing (in purple and red, respectively).

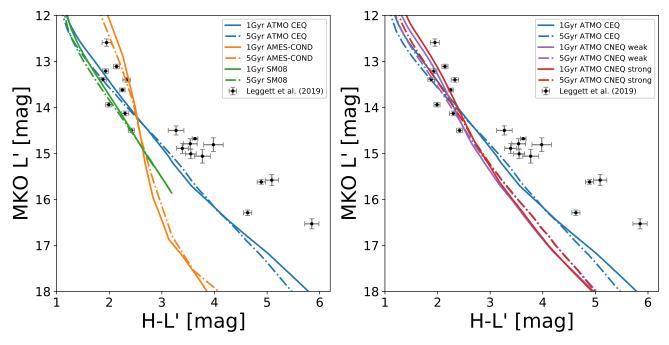


Fig. 14. H-L' colour as a function of absolute L' band magnitude, with photometry of late T and Y dwarfs from Leggett et al. (2019) plotted as black circles. Overplotted in the *left panel* are isochrones of chemical equilibrium models from this work (in blue), from B03 (in orange), and from SM08 (in green.) Overplotted in the *right panel* are isochrones of chemical equilibrium models (in blue), and non-equilibrium models with weak and strong vertical mixing (in purple and red, respectively).

strength of vertical mixing is overestimated in our current model set-up. Decreasing the eddy diffusion coefficient K_{zz} further may improve the comparison to the photometric points in the $4-5 \mu m$ flux window for this object.

Observations of the Y0- and Y1-type objects WISE 1206 and WISE 1541 (Cushing et al. 2011; Schneider et al. 2015) are shown in the middle and bottom panels of Fig. 18, respectively. Using the cloud-less models of SM08, Schneider et al.

(2015) estimate $T_{\rm eff} \sim 400-450\,\rm K$ and $\log(g) = 4.0-4.5$ for the Y0 object WISE 1206. Zalesky et al. (2019) ran retrieval analysis on a sample of Y dwarfs including WISE 1541, retrieving $T_{\rm eff} \sim 325\,\rm K\,\log(g) \sim 5.0$ for this object, in line with comparisons to cloud-free forward models presented in Leggett et al. (2013). Here we compare $T_{\rm eff} = 420\,\rm K$, $\log(g) = 4.5$ models to WISE 1206 and $T_{\rm eff} = 330\,\rm K$, $\log(g) = 4.0$ models to WISE 1541. We use a lower value of the surface gravity for WISE 1541 than

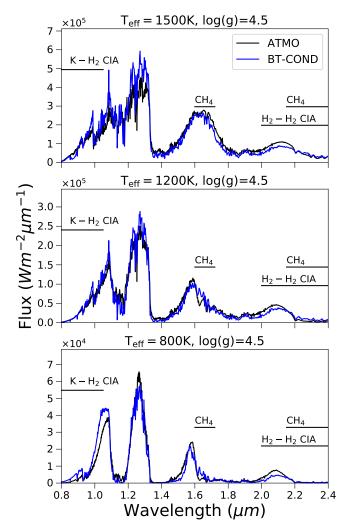


Fig. 15. Synthetic near-infrared emission spectra from ATMO compared with models from the BT-Cond grid (Allard et al. 2012) for a range of effective temperatures.

obtained by previous studies since log(g) = 5.0 does not agree with our evolutionary tracks at this T_{eff} .

The *J*- and *H*-band brightness and shape is better reproduced by the non-equilibrium models for both objects. This is due to the quenching of NH₃ reducing the opacity in these bands (see Fig. 10 and Tremblin et al. 2015). The strong mixing non-equilibrium chemistry model overpredicts the CO absorption in the $4-5\,\mu m$ flux window for the warmer WISE 1206 object, while the weaker mixing model better reproduces the photometric points in this wavelength range. The strong and weak mixing non-equilibrium chemistry models both overpredict the CO absorption in the cooler WISE 1541 object, with the equilibrium model better reproducing the WISE and *Spitzer* photometry.

The Y-band flux at $\sim 1 \, \mu \text{m}$ is underpredicted for WISE 1206 and WISE 1541 by the current models, an issue also seen by the model comparisons in Schneider et al. (2015). We note that the K-H₂ opacity is important in this wavelength region; however, given the improvements to the K resonant line broadening outlined in Sect. 3, we do not attribute this discrepancy to shortcomings in the K opacity. Instead, we note that reducing the K abundance by approximately an order of magnitude rectifies the difference between model and data in the Y band. We therefore posit that the current modelling of the potassium chemistry,

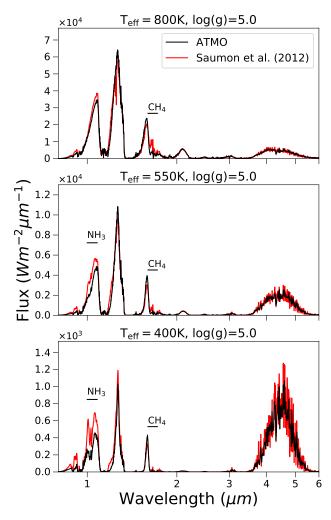


Fig. 16. Synthetic infrared emission spectra from ATMO compared with models from Saumon et al. (2012) for a range of effective temperatures.

including potentially its condensation into KCl and/or the thermochemical data used to calculate the equilibrium abundances, is slightly incorrect. This should be investigated more thoroughly in future work.

5. Discussion and conclusions

We have presented our new ATMO 2020 set of substellar atmosphere and evolutionary models applicable to cool brown dwarfs and directly imaged giant exoplanets. Our atmosphere model grid is generated with our 1D code ATMO, and spans $T_{\rm eff} = 200-3000~{\rm K}$, $\log(g) = 2.5-5.5$ with both equilibrium and non-equilibrium chemistry due to different strengths of vertical mixing. This grid of atmosphere models has been used as the surface boundary condition for the interior structure model to calculate the evolution of $0.001-0.075~M_{\odot}$ objects. We have highlighted numerous theoretical modelling improvements through comparisons to other model sets in the literature and comparisons to observational datasets in Sect. 4. We now discuss and conclude our work.

5.1. Evolutionary tracks

Our key result is that there are notable changes to the cooling tracks of substellar objects over previous works by B03 and SM08 (Fig. 7). These changes are brought about by two major

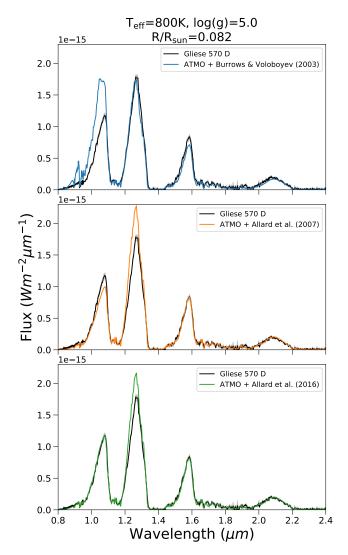


Fig. 17. Model comparisons to the absolutely flux calibrated near-infrared spectrum of the T7.5 dwarf Gliese 570 D (Burgasser et al. 2000). Three models are shown all calculated self-consistently using K resonance line broadening from BV03 (*top*), A07 (*middle*), and A16 (*bottom*), for $T_{\rm eff} = 800 \, {\rm K}$, $\log(g) = 5.0$, and $R/R_{\odot} = 0.082$.

modelling improvements. First, the use of a new hydrogen and helium EOS (Chabrier et al. 2019) in the interior structure model has raised the hydrogen and deuterium minimum masses, causing changes in the cooling tracks around these masses (Fig. 5). This new EOS includes quantum molecular dynamics calculations and predicts cooler, denser, and hence more degenerate objects than the previous semi-analytic EOS of Saumon et al. (1995) used in previous works (Fig. 6). We are planing to work on further improvements to the EOS, in particular improvements concerning the interaction between H and He, which could further alter the cooling curves.

The second major improvement concerns the line lists used for important molecular opacity sources such as CH_4 and NH_3 used in the model atmospheres. These line lists are taken primarily from the ExoMol group (Tennyson & Yurchenko 2018), and include significantly more transitions required to accurately model the high-temperature atmospheres of brown dwarfs and giant exoplanets. These more complete line lists have added opacity to the 1D model atmosphere, changing the predicted emission spectra (Figs. 15 and 16) and leading to warmer temperature structures (Fig. 4). These warmer temperature structures

are used as surface boundary conditions for the interior structure evolution model, and have changed the shape of the cooling tracks of substellar objects in the low-mass brown dwarf regime (Fig. 7). For example, a 1 $M_{\rm Jup}$ object is cooler and less luminous at ages <0.1 Gyr, and warmer and brighter for ages >0.1 Gyr compared to the previous AMES-Cond models. We note that line lists of important molecular absorbers are being continuously updated and improved, and we are working on integrating new $\rm H_2O$ (Polyansky et al. 2018) and TiO (McKemmish et al. 2019) line lists into our 1D atmosphere model. Investigating the impact of this updated opacity on the temperature structures, synthetic spectra, and cooling tracks will be the subject of future work.

5.2. Potassium broadening

Along with improvements to the molecular opacities, we have highlighted the improvement and importance of the pressure broadened potassium resonance doublet in Sect. 3. We have implemented new K line shapes of A16 in our 1D atmosphere model ATMO. These new line shapes improve upon the line shapes of A07, including better determinations of the intermediate- and long-range part of the K-H₂ interaction potential and spin-orbit coupling. We compare synthetic emission spectra calculated with these new K line shapes to others commonly used in the literature (BV03; A07), with our key finding being that there is a large impact and uncertainty on the predicted Y- and J-band flux due to these different line shapes. The large differences in opacity in the K far-red wing alters the flux through the Y band at $\sim 1 \,\mu\text{m}$, and leads to a redistribution of flux to longer wavelengths due to differences in the P-T profile, primarily impacting the J band. This redistribution of flux is only captured when reconverging the P-T profile to find radiative-convective equilibrium when switching between opacity sources.

To validate the new K line shapes of A16, we compare spectra computed with different K broadening line shapes to a typical late T dwarf Gliese 570 D (Fig. 17). We find that the Y band at $\sim 1 \,\mu\text{m}$ is best reproduced by models including the new line shapes, with models computed with A07 and BV03 line shapes predicting too strong and too little absorption, respectively, in the far-red wing. Further work must be undertaken to validate these line shapes not only in the near-infrared, but also at redoptical wavelengths where these line shape calculations differ in the blue wing of the doublet (see Figs. 2 and 3). The blue wing displays a satellite feature brought about by K-H₂ quasimolecular absorption that has previously been detected in the T dwarf ϵ Indi Ba (Allard et al. 2007a), and can be a useful diagnostic of temperature and metallicity (A07). Accurately modelling the optical spectrum requires taking into account the pressure broadened line shapes of other alkali metals such as Na and Li. Recently, Allard et al. (2019) presented improvements on the line shapes of the Na resonance doublet, finding a change in the blue wing of the doublet in the predicted emission spectra of self-luminous atmospheres. We are working to include these new Na resonance line shapes in our 1D atmosphere code, and future work to validate these improvements on the alkali opacity at red optical wavelengths must be undertaken.

5.3. Non-equilibrium chemistry

In our new model set we have considered two chemistry scenarios, thermodynamic equilibrium and non-equilibrium due to vertical mixing. To model non-equilibrium chemistry we have consistently coupled the relaxation scheme of Tsai et al. (2018) to our 1D atmosphere code, considering the non-equilibrium abundances of H₂O, CO, CO₂, CH₄, N₂, and NH₃. We adopt this

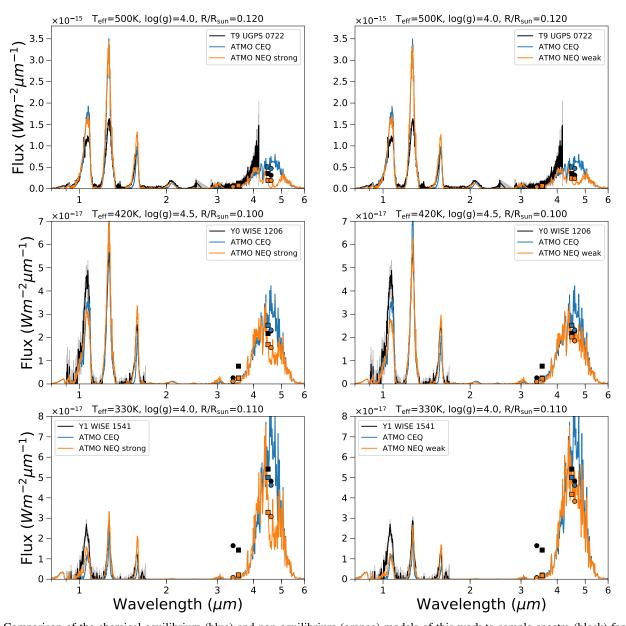


Fig. 18. Comparison of the chemical equilibrium (blue) and non-equilibrium (orange) models of this work to sample spectra (black) forming a T–Y spectral sequence. *Left and right columns*: non-equilibrium models with the strong and weak K_{zz} mixing relationships with surface gravity, respectively. *Spitzer* IRAC photometry is plotted as squares and WISE photometry as circles. *Top panel*: observed spectrum of the T9 dwarf UGPS 0722 (Lucas et al. 2010; Leggett et al. 2012), with *Spitzer* IRAC and WISE photometric points from Kirkpatrick et al. (2012). *Middle and bottom panels*: 0.9–1.7 μ m HST WFC3 spectra of the Y0- and Y1-type dwarfs WISE 1206 and WISE 1541 (Schneider et al. 2015), with *Spitzer* IRAC photometry also from Schneider et al. (2015) and WISE photometry from Cutri et al. (2013).

relaxation scheme over a full kinetics network for computational efficiency and consistent convergence throughout the grid when solving for a self-consistent P–T profile. This relaxation scheme is more computationally efficient as it avoids the need to solve the large, stiff system of ordinary differential equations needed when using full chemical kinetics networks.

While we have considered the non-equilibrium abundances of the primary carbon- and nitrogen-bearing molecules, future models should include additional species thought to be impacted by vertical mixing. Non-equilibrium signatures of HCN may become apparent in high-gravity objects with vigorous mixing (Zahnle & Marley 2014), and PH $_3$ and GeH $_4$, both of which are signatures of vertical mixing in Jupiter's atmosphere, as well as C_2H_2 and CH_3D , could impact the mid-infrared spectra of the coolest brown dwarfs (Morley et al. 2018). Furthermore, present

chemical kinetics models do not consider condensate species. As such, the models essentially assume that mixing of species into the upper atmosphere happens on timescales much shorter than condensation timescales. Such an assumption is important for $\rm H_2O$ and $\rm NH_3$, which condense in the upper atmospheres of cool brown dwarfs in chemical equilibrium. Incorporating condensation timescales would involve combining kinetic cloud formation models such as the Helling & Woitke model (Woitke & Helling 2003, 2004; Helling & Woitke 2006; Helling et al. 2008) with a gas-phase chemical kinetics scheme. While this coupling is technically challenging and beyond the scope of this work, coupled gas-cloud kinetics models are required to correctly determine the abundances of $\rm H_2O$ and $\rm NH_3$, which are critical species governing the temperature structure and thermal emission from cool Y dwarf atmospheres.

We demonstrate the impact of nitrogen and carbon nonequilibrium chemistry due to vertical mixing on the predicted emission spectra of cool T-Y-type objects (Fig. 10). The quenching of CH₄ and NH₃ brightens the H band, and similarly to Hubeny & Burrows (2007) we find the increased abundances and absorption of CO and CO₂ suppress the flux in the $3.5-5.5 \mu m$ flux window. The comparisons to the observed spectra of cool brown dwarfs in Fig. 18 indicate the quenching of ammonia is key in reproducing near-infrared emission of Y dwarfs, in agreement with the results of Leggett et al. (2015) and Tremblin et al. (2015). Comparisons of our new models to the H-W2 colours of cool brown dwarfs (Fig. 13) and to photometric observations across the T-Y sequence (Fig. 18) also support the presence of non-equilibrium CO and CO₂ abundances in the $3.5-5.5 \mu m$ flux window. These results indicate that this wavelength region could be a useful tool in constraining the eddy diffusion coefficient K_{72} , the parameter commonly used to model vertical mixing in 1D atmosphere codes. The value of K_{zz} in brown dwarf and exoplanet atmospheres is a long-standing theoretical problem, and several attempts have been made to estimate it from numerical models (e.g. Moses et al. 2011; Parmentier et al. 2013; Zhang & Showman 2018a,b). Since the abundances of CO and CO₂ depend strongly on the value of K_{zz} , the 3.5-5.5 μ m flux window could be a useful observational diagnostic of K_{zz} in cool brown dwarf atmospheres.

While non-equilibrium chemistry can reproduce several observational features of cool brown dwarfs, the red colours of late T dwarfs cannot be reproduced by cloud-free equilibrium or non-equilibrium models, similarly to the red late L dwarfs and the L–T transition. This is demonstrated by our comparisons to *J–H* colours in colour-magnitude diagrams (Fig. 11), and our spectral comparisons to the late T dwarf UGPS 0722 in Fig. 18. Developing models which can reproduce this observed reddening will be the subject of future work.

5.4. Future work

Over the past half decade, a new theory has been developed suggesting that chemical transitions such as $CO \rightarrow CH_4$ and $N_2 \rightarrow NH_3$ in brown dwarf atmospheres can be responsible for triggering convective instabilities. This can reduce the temperature gradient in the atmosphere reddening the emission spectrum. Reductions in the temperature gradient through the effective adiabatic index γ_{eff} have been shown to reproduce several observed features of brown dwarfs, including the L–T transition (Tremblin et al. 2016), extremely red young low-gravity objects (Tremblin et al. 2017a), and the red colours of cool late T dwarf objects (Tremblin et al. 2015).

To investigate the potential mechanism at play reducing the temperature gradient in brown dwarf atmospheres, Tremblin et al. (2019) generalised convection and mixing length theory to include diabatic processes through thermal and compositional source terms, demonstrating that a number of convective systems in the Earth's atmosphere and oceans derive from the same instability criterion. In brown dwarf atmospheres, the thermal and compositional source terms are represented by radiative transfer and $CO \rightarrow CH_4$ or $N_2 \rightarrow NH_3$ chemistry, respectively, with the convective instability driven by opacity and/or mean molecular weight differences in the different chemical states. The idealised 2D hydrodynamic simulations of Tremblin et al. (2019) reveal that by including such source terms the temperature gradient can indeed be reduced to that required to qualitatively reproduce brown dwarf observations. Motivated by this, we aim to expand on this initial grid of atmosphere models by reducing

the temperature gradient through the effective adiabatic index $\gamma_{\rm eff}$. Our key goal is to further study how radiative convection may evolve and influence the brown dwarf cooling sequence. We look forward to the good-quality, wide spectral coverage observational datasets that future instrumentation such as the *James Webb* Space Telescope will provide to aid in such studies.

5.5. ATMO 2020 publicly available models

The models presented in this work are publicly available for download^{4,5}. This includes P–T profiles, chemical abundances, and top of the atmosphere emission spectra, as well as evolutionary tracks with absolute magnitudes in a number of common photometric filters. We encourage users to contact the author if additional models or photometric filters are required, and while these are currently solar metallicity models, we plan to generate non-solar metallicity models in future work.

It should be noted that this grid is focused on cool T–Y-type brown dwarfs and that the range of validity of our models is $T_{\rm eff} \lesssim 2000~\rm K$, due to the calculations missing some sources of opacity (e.g. some hydrides, condensates) that form at higher temperatures and/or the modified atmosphere thermal gradients suggested for L dwarfs (Tremblin et al. 2016, 2017b). Such issues will be addressed in future studies. Therefore, for brown dwarfs with $T_{\rm eff} \gtrsim 2000~\rm K$ (i.e. essentially massive objects younger than $100~\rm Myr$ with $M > 0.015~M_{\odot}$), we suggest using either the Chabrier et al. (2000) or Baraffe et al. (2015) models. We note that the latter include an updated treatment of atmospheric convection.

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⁴ http://opendata.erc-atmo.eu

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