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Weak- and Strong-Friction Limits of Parcel Models: Comparisons and Stochastic Convective Initiation Time[†]

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For moist convection, models of individual parcel dynamics are valuable for their simple formulation and predictions of cloud properties. Here, two limiting idealized cases of parcel theory are investigated: the weak- and strong-friction limits. The weak-friction limit is a traditional limit with no momentum drag, and the dynamics are a Hamiltonian system for the parcel's height and vertical velocity. A strong-friction limit is derived and studied here, and its limiting form involves a balance between frictional drag and buoyancy, which provides a differential equation for parcel height as a function of time. In the two limiting regimes, analytical formulas are presented and compared for quantities such as maximum vertical velocity and cloud-top height. For example, in the strong-friction limit, the cloud-top height coincides with the level of neutral buoyancy (LNB), whereas in the weak-friction limit the cloud-top height is far above the LNB. This comparison suggests the strong-friction limit may provide more realistic predictions of some averaged cloud properties. In general, since frictional effects and individual parcel properties can vary even within a single cloud, the predictions of the weak- and strong-friction limits can be viewed as upper and lower bounds for the behavior of more realistic finite-friction scenarios. Finally, in a stochastic version of the parcel model, in the strong-friction limit, analytical formulas are derived for convective initiation time. Applications to convective parameterizations are discussed; for example, the formulas for convective initiation time could be applied as stochastic convective triggers.

Key Words: Parcel models, plume models, stochastic dynamics, adiabatic parcel

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50 1. Introduction

51 Moist convection and cloud dynamics are complex processes, and 52 simple theories have long been sought after. In one simplified 53 perspective, moist convection is viewed in terms of fluid parcels 54 (e.g., Simpson and Wiggert 1969; Brast et al. 2016). In the most 55 idealized scenario, each parcel rises as an undiluted, undamped 56 entity, with its vertical velocity generated through the parcel's 57 buoyancy. Since this idealized scenario neglects momentum 58 damping, we will call it the weak-friction limit. 59

In this paper, the main goal is to derive and investigate the dynamics of the opposite limit: the strong-friction limit. In this limit, the parcel's vertical motion is strongly damped, and the vertical acceleration is relatively small, which leaves a dominant balance between frictional drag $(\tau_w^{-1}w)$, buoyancy (b), and other forcing (f):

$$\frac{1}{\tau_w}\frac{dz}{dt} = b(z) + f.$$
(1)

(If the frictional drag is parameterized in a form other than $\tau_w^{-1}w$, then a similar balance and similar differential equation for height z(t) could be derived, although its detailed form would be different.)

The overarching questions of the present paper are: Can analytical formulas be derived in the strong-friction limit? Based on the analytical formulas, what are the cloud and parcel properties that are predicted in the strong-friction limit? How do the predictions of the strong-friction limit compare with predictions of the weak-friction limit? Analytical formulas are one

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of the valuable aspects of the weak-friction limit, and analytical formulas for the strong-friction limit could potentially provide the idealized counterpart of the well-known weak-friction limiting dynamics.

5 Some interesting studies have used large eddy simulations to 6 investigate the strength of frictional drag on fluid parcels. Some 7 results suggest that friction may be weak, a scenario they call "slippery thermals" (Sherwood et al. 2013). Other results suggest 8 that friction may be strong, a scenario they call "sticky thermals" 9 (Romps and Charn 2015). Nature may perhaps involve a complex 10 variety of scenarios (Hernandez-Deckers and Sherwood 2016). In 11 the present study, the goal is not to assess the strength of frictional 12 drag, but to examine the idealized parcel models that arise in the 13 two extreme cases of the weak- and strong-friction limits. 14

As one main theme here, the properties of an adiabatic, rising 15 parcel are derived in the strong-friction limit. For example, the 16 vertical velocity profile can be written analytically in terms 17 of parameters such as the environmental moist thermodynamic 18 profile. Compared to the traditional weak-friction limit, an 19 adiabatic strong-friction parcel reaches its maximum vertical 20 velocity at a much lower altitude. One would expect the weak-21 and strong-friction limits to be useful for upper and lower bounds 22 on predictions of cloud properties, with the more realistic finite-23 friction dynamics lying in between.

24 Another main quantity of interest here will be convective 25 initiation time, the time elapsed in waiting for a boundary-layer 26 parcel to reach a barrier located above in the vicinity of the level of 27 free convection. In the strong-friction limit, in a stochastic version 28 of the model, an approximation of the mean convective initiation time can be found in analytic form. Such an analytic formula could 29 be applied in convective parameterizations as a simple stochastic 30 trigger (Kain and Fritsch 1992; Lin and Neelin 2000; Majda and 31 Khouider 2002; Jakob and Siebesma 2003; Stechmann and Neelin 32 2011, 2014; Gentine et al. 2013a,b; D'Andrea et al. 2014; Hottovy 33 and Stechmann 2015). Indeed, one of the main applications of 34 parcel and plume models is to convective parameterizations. 35

The rest of the paper is organized as follows. The parcel model 36 for convection is described in section 2, and some motivating 37 examples of parcel pathways are illustrated in section 3. Then, 38 in sections 4 and 5, derivations of the weak- and strong-friction 39 limits are described, and their parcel properties are compared 40 in terms of vertical velocity, cloud-top height, etc. Convective 41 initiation time is investigated in a stochastic version of the 42 parcel model in section 6, including approximations that allow 43 for analytical formulas. Finally, sections 7 and 8 present some 44 additional discussion and conclusions. 45

2. Model description

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As a simple model of moist convection, the dynamics of a parcel will be given by the stochastic differential equation (SDE)

$$dz = w dt,$$

$$dw = b dt - \frac{1}{\tau_w} w dt + b_w dW.$$
(2)

54 Here, z and w are the parcel's height and vertical velocity at 55 time t, respectively. The fluid dynamical interactions between 56 the parcel and the rest of the atmosphere is incorporated with a 57 simple relaxation in the vertical velocity $(-w/\tau_w)$ and a stochastic 58 term $(b_w dW)$. Physically, one contributor to the relaxation could 59 be entrainment-the mixing process between the parcel and 60 the environment-although other significant contributions could come from in-cloud fluid dynamical effects. Here τ_w is the relaxation timescale for the vertical velocity, which is relaxed toward the zero vertical velocity of the parcel's surroundings. Intuitively, and in brief, dW is a random forcing. The infinitesimal

Wiener increment dW is the limit of the small increment $\Delta W = W(t + \Delta t) - W(t)$, in the limit as $\Delta t \rightarrow 0$, in much the same way that dz is the limit of $\Delta z = z(t + \Delta t) - z(t)$. For finite Δt , a new value of ΔW is chosen at each time step, and each value is an independent sample from a Gaussian distribution with mean 0 and variance Δt . In the limit of $\Delta t \rightarrow 0$, the SDE is obtained and it is interpreted here in the Ito sense (Gardiner 2004). Furthermore, the buoyancy b is given by

$$b = \frac{g}{\theta_o} \left(\theta_{\rm v} - \theta_{\rm v,env} \right), \tag{3}$$

where $g = 9.81 \text{ m s}^{-2}$ is the acceleration of gravity, $\theta_o = 300 \text{ K}$ is a reference value of the potential temperature at sea surface, and θ_v and $\theta_{v,env}$ are the virtual potential temperatures of the parcel and environment, respectively, which are to be specified further below in Section 2.1.

Note that the parcel's thermodynamic properties are assumed to be conserved – i.e., it is an adiabatic parcel model. While such an assumption may not be realistic, it is a convenient simplification that is commonly used (e.g., Xu and Emanuel 1989; Emanuel 1994; Grabowski and Jarecka 2015) and allows for the derivation of some analytical formulas. The possibility of modeling nonadiabatic parcel dynamics is discussed below in section 7.

One difference from typical parcel models is that the model here is time dependent. For instance, other stochastic parcel models have been considered in the past (e.g., Fraedrich 1985), although without time dependence. A time-dependent model is needed for the analysis of certain quantities of interest, such as the time it takes for a parcel to reach a barrier located above in the vicinity of the level of free convection (convective initiation time), and we will show below that explicit approximations can be obtained under certain conditions.

2.1. Simplified thermodynamic relations

The model in (2)–(3) can be closed by defining both the environment- and parcel- virtual potential temperatures, $\theta_{v,parcel}$ and $\theta_{v,env}$. To this end, we will use linearized versions of the full thermodynamic relations (Majda and Xing 2010; Deng *et al.* 2012; Hernandez-Duenas *et al.* 2013). Such relations simplify the model significantly compared to more comprehensive thermodynamics relations, since they allow explicit formulas to be obtained in many situations, as seen below. However, we have also investigated more comprehensive thermodynamic relations, and the main results of this paper are essentially the same, although some results can then only be obtained numerically, not analytically, so we instead focus here on the simplified, linearized thermodynamics.

In this spirit, the virtual potential temperature is given by

$$\theta_{\rm v} = \theta_{\rm e} + \theta_o \left(\epsilon_o - \frac{L}{c_p \theta_o}\right) q_{\rm v} - \theta_o q_\ell. \tag{4}$$

Here θ_e is the equivalent potential temperature, q_v, q_ℓ the water vapor and liquid water mixing ratios, $c_p = 10^3 \text{J kg}^{-1} \text{K}^{-1}$ the specific heat at constant pressure, $L = 2.5 \times 10^6 \text{ J kg}^{-1}$ is the latent heat of vaporization, and $R_v/R_d = 1 + \epsilon_o$ is the ratio of gas constants for water vapor (R_v) and dry air (R_d) with $\epsilon_o \approx$ 0.6. This formula can be used, as described below, to define the parcel and environmental θ_v values by inserting parcel and environmental values, respectively, into the right-hand side.

The parcel will be assumed to be adiabatic in the sense that its total water q_t and equivalent potential temperature θ_e are assumed to be conserved and not changing in time. We will assume that there is neither precipitation nor formation of rain. All the water that is condensed is kept with the parcel, leaving

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Table 1. Simplified thermodynamic relations.

Env. quantities	
$\theta_o = 300K,$	$\theta_{\rm env}(z) = \theta_o + Bz$
$B = 3 \mathrm{K} \mathrm{km}^{-1}$	
$q_{ m v,env}(z)$	$\theta_{e,env}(z) = \theta_{env}(z) + \frac{L}{c_p}q_{v,env}(z)$
	$\theta_{\rm v,env}(z) = \theta_{e,{\rm env}(z)}$
	$+ heta_o\left(\epsilon_o-rac{L}{c_p heta_o} ight)q_{ m v,env}(z)$
Parcel quantities	Definition
z, w	Height and vertical velocity.
θ_e	Equivalent pot. temperature.
q_t	Total water mixing ratio.
Relations	Comments:
p(z)	Total pressure as a function
	of height only.
$q_{ m vs}(z)$	Water vapor at saturation as a
	function of height only.
$q_v = \min(q_t, q_{\rm vs})$	Water vapor mixing ratio.
$q_{\ell} = \max(q_t - q_{\rm vs}, 0)$	Liquid water mixing ratio.
$\theta_{\rm v} = \theta_{\rm e} - \theta_o q_\ell$	Virtual potential temperature
$+ heta_o\left(\epsilon_o-rac{L}{c_p heta_o} ight)q_{ m v}$	
$b = \frac{g}{\theta} (\theta_v - \theta_{v,env})$	Buoyancy

27 the total water $q_t = q_v + q_\ell$ constant, as in a reversible process. 28 The presence of rain fallout would increase the parcel's virtual 29 potential temperature and increases the cloud top height. We have explored those situations with different rainfall speeds in 30 the model (not shown) and no qualitative differences have been 31 observed other than the one mentioned above. In the spirit of 32 a simplified setting, we then only consider two constituents of 33 water: water vapor and cloud liquid water. As part of the simplified 34 thermodynamics, we also assume a saturation water vapor $q_{ys}(z)$ 35 as a function of height only. The dependence on z is chosen to be 36 a function with roughly exponential decay, as used in Hernandez-37 Duenas et al. (2013). The water vapor and liquid water are then 38 given by

$$q_v = \min(q_t, q_{vs}(z)), q_\ell = \max(q_t - q_{vs}(z), 0).$$
(5)

We have also explored other choices. For instance, extracting all the condensed water instantaneously from the parcel can be interpreted as having a vanishing liquid water, as in a pseudoadiabatic process. We note that such a scenario results in a slightly higher cloud top height. However, there is no qualitative differences compared to the previous case.

Plots of the environmental thermodynamic profiles are shown here in Fig. 1. The environmental potential temperature $\theta_{\text{env}}(z) = \theta_o + Bz$ is taken to be linear with respect to height with rate $B = 3 \text{ K km}^{-1}$. With these choices, the environmental -equivalent and -virtual potential temperature are

$$heta_{
m e,env}(z) = heta_{
m env}(z) + rac{L}{c_p}q_{
m v,env}(z),$$

$$heta_{ ext{v,env}}(z) = heta_{ ext{e,env}}(z) + heta_o \left(\epsilon_o - rac{L}{c_p heta_o}
ight) q_{ ext{v,env}}(z).$$

(6)

Table 1 includes all the variables and the simplified thermodynamic relations.

59 The virtual potential temperature fluctuation, $\theta_v - \theta_{v,env}$, is then 60 an explicit function of height, given by

$$\theta'_{v}(z) = \theta_{e} - \theta_{e,env}(z) + \theta_{o} \left(\epsilon_{o} - \frac{L}{c_{p}\theta_{o}} \right) (q_{v}(z) - q_{v,env}(z)) - \theta_{o}q_{\ell}(z),$$
(7)

and the buoyancy from (3) can likewise now be defined accordingly as an explicit function of height:

$$b(z) = \frac{g}{\theta_{\star}} \theta'_{\nu}(z). \tag{8}$$

This completes the specification of the model.



Figure 1. Top: Mixing ratios showing total water (dashed curve), environmental water vapor (solid line) and water vapor at saturation (dots). Bottom: Virtual and equivalent potential temperatures for environment and parcel. The symbols for each temperature are specified in the legend.

To define the initial conditions, the parcel is initially located at a height z_o and its thermodynamic properties are assumed to be in balance with its environment, i.e., the equivalent potential temperature and moisture of the parcel and the environment coincide $\theta_e = \theta_{e,\text{env}}(z_o), q_v = q_{v,\text{env}}(z_o)$. As a consequence, the virtual potential temperatures are also initially in balance $\theta_v =$ $\theta_{v,\text{env}}(z_o)$. Here we choose this initial height z_o to be close to the surface: $z_o = 0.5$ km. Fig. 1 shows the profiles for water vapor (top panel) and the equivalent and virtual potential temperatures (bottom panel) where one can corroborate the intersection of the profiles at the initial parcel's height.

For a rising parcel, three heights are particularly important, and while they are well-known, it is useful to briefly summarize them here to facilitate further discussions below. Fig. 1 (bottom panel) shows the three heights, and it also illustrates that the three heights mark transitions between positive and negative buoyancy. In particular, as a parcel first rises from the boundary layer, its virtual potential temperature is conserved up until the *lifting condensation level* (LCL), which is the level where the parcel's water vapor starts to condense (see the bottom panel

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of Fig. 1). Since the environmental virtual potential temperature increased during all this time, the parcel becomes negatively buoyant. However, if the parcel continues to rise above the lifting condensation level, its virtual potential temperature increases and it could intersect with the environmental virtual potential temperature again. The height where it occurs is known as the level of free convection (LFC). The parcel becomes positively buoyant again above the LFC and can continue rising without any other forcing assuming it was able to reach such a height. Finally, the buoyancy vanishes again at the so called level of neutral 10 buoyancy (LNB). The level of neutral buoyancy is sometimes 11 related to the cloud top height. In reality, it is the place where 12 the parcel starts decelerating due to buoyancy. Depending on the 13 properties of the system (like mixing with the environment), LNB 14 could provide one estimate for the cloud top height or the parcel 15 could go much higher up. We will analyze such situations in more 16 detail in Section 5. 17

2.2. Potential energy and total energy

It will be useful to define a total energy E of the system, and to relate it to the concepts of CIN and CAPE (to be described below). The total energy is defined as

$$E(z,w) = \frac{1}{2}w^2 + \Pi(z),$$
(9)

which is an explicit function of height and vertical velocity. The first term, $w^2/2$, is the kinetic energy, and the second term,

$$\Pi(z) = \frac{g}{\theta_o} \int_z^H \theta'_v(z') dz' \tag{10}$$

is the potential energy of the system. Here H = 15 km is the 32 troposphere's depth. The mass m of the parcel is here assumed 33 to be constant. We note that we have re-scaled the actual energy 34 $\frac{1}{2}mw^2 + m\Pi(z)$ by the constant mass m (see Section 4.1 for 35 further discussion). 36

The concepts of convective inhibition (CIN) and convective 37 available potential energy (CAPE) are related to the potential 38 energy $\Pi(z)$, as illustrated in the bottom panel of Fig. 2. The 39 figure shows that $\Pi(z)$ is a double well potential. The bottom well 40 has a local minimum at the parcel's initial height, z_o . The upper 41 well has a local minimum at the LNB. In between the two wells 42 is a metastable local maximum (a saddle point) at the LFC. The 43 convective inhibition

$$CIN = \Pi(z_o) - \Pi(LFC) \tag{11}$$

is a measure of the depth of the bottom well, and it is defined as the 47 potential energy difference between the bottom of the well and the 48 level of free convection. It is negative and is associated with the 49 amount of potential energy that the parcel would need to convert 50 to kinetic energy in order to escape from the bottom well. On the 51 other hand, the convective available potential energy 52

$$CAPE = \Pi(LFC) - \Pi(LNB)$$
(12)

55 is a measure of the depth of the upper well, and it is defined as 56 the potential energy difference between the bottom of the upper 57 well and the saddle point. In the plot, vertical dashed lines were 58 included to identify CIN and CAPE more clearly. The parameter values that we chose give $CIN = -152.68 \text{ J kg}^{-1}$ and CAPE =59 1214.2 J kg⁻¹. A value of |CIN| greater than 200 Jkg⁻¹ can 60 be sufficient to prevent convection in the atmosphere. The value of CAPE reported here is about one quarter the values that are reported for extreme events. We also note that the upper well is much deeper than the bottom one. This depth's difference has



Figure 2. Top: Energy contours (thin solid curve) and separatrix (thick solid curve) E = E(w = 0, z = LFC). Red and blue arrows (labeled for the printed version) show an schematic of preferred transition paths for the strong and weak friction regimes respectively (to be explained below). Bottom: Potential energy, showing CIN (CAPE) as the potential energy difference between z_o (LFC) and LFC (LNB). The location of the barrier (LFC⁺) above in the vicinity of LFC also shown.

implications for the parcel's dynamics when stochastic variability is added, as it is explained in the next sections.

Fig. 2 (top panel) shows contours of the corresponding total energy function, E(z, w). It has three critical points, corresponding to the bottoms of two wells and a saddle point that connects them. The saddle point is an metastable equilibrium point, identified with the level of free convection (LFC). The separatrix here is defined as the curve passing through the saddle point dividing the two wells, and it is plotted with the thick solid line. The separatrix has significance in that it encloses a set of closed contours.

The energy is relevant to the main discussion of the paperweak and strong friction limits-in the following way. The terms in the parcel model in (2) can be divided into two groups: a deterministic oscillatory component due to buoyancy (w dt, b dt), and a forced-damped component from the stochastic forcing and mixing of the parcel with its environment $(-wdt/\tau_w + b_w dW)$. When different parameter regimes are explored, one may see the dominance of either of these two components, or possibly both terms equally contributing to the time evolution of the parcels. In particular, we distinguish two main parameter regimes of strong and weak friction. Fig. 2 (top panel) shows an schematic of the observations to be made below with two cartoon arrows on top of the energy contours. It will be seen below that the weak-friction

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1 limit corresponds to an evolution that follows an energy contour 2 and has energy conservation, whereas the strong friction limit 3 corresponds to an evolution that crosses energy contours due to 4 frictional effects. 5

2.3. Parameter calibration

In subsequent sections, the model parameters τ_w and b_w in (2) will take on different values to describe the range of behaviors of the weak- and strong-friction limits. To see the impact of 10 the parameters τ_w and b_w on the behavior of the model, it is 11 useful to consider a simpler version of the model in (2) by 12 neglecting the buoyancy. In that case, w would evolve according 13 to $dw = -(w/\tau_w) dt + b_w dW$, which is known as the Ornstein– 14 Uhlenbeck process (e.g., Gardiner 2004). This equation involves 15 two components: an exponential decay from the friction, dw =16 $-(w/\tau_w) dt$, plus a random forcing $b_w dW$. This equation also 17 arises in other contexts, such as a particle undergoing Brownian 18 motion under the influence of friction. The values for τ_w and 19 b_w are then seen to be associated with two statistics. First, the 20 covariance of w(t) and w(s) at two times t and s is given by 21

$$\frac{\tau_w b_w^2}{2} \left(e^{-|t-s|/\tau_w} - e^{-(t+s)/\tau_w} \right) + \operatorname{var}(w(0)) e^{-(t+s)/\tau_w}.$$
(13)

From this formula, one can see that τ_w describes the decorrelation time. Second, considering a single time s = t, in the stationary limit as $s = t \rightarrow \infty$, one obtains the stationary variance; its square root is

$$\sigma_w = b_w \sqrt{\tau_w/2}.\tag{14}$$

This is the stationary standard deviation, and it is related to both of the parameters τ_w and b_w . These two statistical quantities can therefore be used for guidance below in selecting and interpreting values of the model parameters b_w and τ_w .

3. Pathways through w-z phase space

37 In order to illustrate the parcel dynamics in the weak-friction and 38 strong-friction regimes, we first present numerical solutions of (2) using the Euler-Maruyama (EM) method (e.g., Higham 2001). 39 In brief, the Euler-Maruyama method works essentially the same 40 as the commonly used forward Euler method for deterministic 41 differential equations, except a random forcing is also added 42 to represent dW(t). We refer to one "realization" of (2) as 43 the dynamics of z(t) and w(t) that result for one particular 44 sequence of random forcing values ΔW_0 , ΔW_1 , ΔW_2 , \cdots , at 45 time steps $t = t_0, t_1, t_2, \cdots$. A different realization is obtained by 46 drawing different random values for $\Delta W_0, \Delta W_1, \Delta W_2, \cdots$. Each 47 value of ΔW_n $(n = 0, 1, 2, \dots)$ is an independent sample from a 48 Gaussian distribution with mean 0 and variance Δt . Furthermore, 49 by considering an ensemble of many realizations, the probability 50 density can be estimated for the evolution of the parcel.

51 Fig. 3 shows a case corresponding to weak friction. Here the 52 relaxation timescale is $\tau_w = 7.5 \min$, which is comparable to the 53 value used in Neggers *et al.* (2002). The noise amplitude is $b_w =$ 54 4.91×10^{-1} ms⁻¹ s^{-1/2}. Under an Ornstein-Uhlenbeck process, 55 as mentioned above in section 2.3, such parameters would have 56 a quasi-stationary standard deviation of $\sigma_w = 7.37 \text{ m s}^{-1}$, which 57 is consistent with typical fluctuations of the parcel while in the 58 lower well. The top panel of Fig. 3 shows part of the picture of the 59 evolution of one realization of the parcel from 0 to 5 hours. After 60 first spending considerable time fluctuating in the lower well, the parcel eventually escapes. On its approach toward the upper well, the parcel does not move directly toward the bottom of the well; instead, the parcel ascends along the separatrix with total energy approximately conserved.



Figure 3. Top: A parcel realization given by equation (2). Also included are the separatrix (thin solid line) and energy contours near the upper well (thick solid lines). Here the noise's amplitude is $b_w = 4.91 \times 10^{-1} \text{ ms}^{-1} \text{ s}^{-1/2}$, the relaxation timescale is $\tau_w = 7.5$ min and the ending time is 5 h. Bottom: Contours of the probability density function (p.d.f.) in the (z, w) plane, in logarithmic scale. Also included is the separatrix (thin solid line)

While the top panel of Fig. 3 shows only one realization, one can see the statistics of preferred paths by analyzing a probability density function (p.d.f.). We obtain an approximation of p.d.f.'s by proceeding with 1000 simulations and accounting for the incidence of a parcel in a bin in the (z, w) plane. The p.d.f.'s are in units of $(ms^{-1} km)^{-1}$ and normalized with respect to the L^1 norm. The first thing to note about the p.d.f., shown in the bottom panel of Fig. 3, is that the most probable state is, by far, the location of the bottom of the lower well, near w = 0 and z = 0.5km. This indicates that the convective initiation time is long, and the parcel stays near the bottom well for a long time. As a result, contours of the p.d.f. in other regions of phase space might be hard to visualize, so a logarithmic scale is used for the contours of the p.d.f. We also show a mean path in the following sense. A target height z = 8 km is chosen. For each realization we verify whether or not the parcel reaches that height in less than 5 h (the duration of the simulations). If so, we compute the instant of time when the parcel reaches the target and save the path in the (z, w) plane for the previous 15 min. We average over all paths that reached that target. The darker shades indicate the preferred path and aligns with the averaged path. We note that the parcel first accelerates and ascends following a curve close to an energy contour. In a sense, the deterministic component of the system dominates in this case with weak friction.

The statistically preferred path can be quite different depending on various aspects such as environmental profiles and mixing

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Figure 4. Top: Same as in Fig. 3, with stronger relaxation $\tau_w = 0.75 \text{ min}$ and stronger noise amplitude $b_w = 1.55 \text{ms}^{-1} \text{s}^{-1/2}$ in the vertical velocity. Bottom: Contours of the probability density function (p.d.f.) in the (z, w) plane, in logarithmic scale. Also included are the separatrix (thin solid line).

38 of the parcel with its environment among others. As a second 39 example, the top panel of Fig. 4 repeats the simulation done 40 in Fig. 3 with a stronger friction (smaller τ_w) and noise amplitude in the vertical velocity (larger b_w). The timescale here 41 is $\tau_w = 0.75 \text{ min}$ and the noise coefficient is chosen as $b_w =$ 42 $1.55 \text{ms}^{-1} \text{s}^{-1/2}$. We note that τ_w was reduced by a factor of 10 43 while b_w was increased by a factor of $\sqrt{10}$ to maintain the same 44 quasi-steady standard deviation $\sigma_w = 7.37 \text{ ms}^{-1}$. In this case, 45 Fig. 4 shows a quite different behavior. Here, when the parcel 46 escapes from the lower well, it transitions to the upper well in 47 a completely different way. Instead of closely following energy 48 contours, the parcel crosses over them and moves almost directly 49 toward the bottom of the upper well. This behavior is indicative of 50 a more significant contribution of the stochastic component and 51 the strong friction.

52 A schematic summary of Figures 3 and 4 was shown earlier 53 in Fig. 2. In brief, in the weak-friction regime, the parcel appears 54 to rise along a Hamiltonian contour (the separatrix); on the other 55 hand, in the strong-friction regime, the parcel appears to cross 56 Hamiltonian contours as it rises somewhat directly toward the 57 bottom of the upper well. In the strong-friction regime, one could 58 say that the parcel escapes the lower well by moving across the 59 saddle point. 60

4. Weak- and strong-friction limits

In the previous section, two numerical examples were presented to illustrate the different behavior of parcels in the weak- and strong-friction regimes. In the present section, we seek simplified equations to describe the dynamics in these two limiting regimes. In particular, we show that the weak- and strong-friction limits can be described by reduced equations that arise from asymptotic analysis. Furthermore, a critical friction strength is also identified as the crossover point between the weak- and strong-friction regimes.

4.1. Weak-friction limit and Hamiltonian system perspective

In a weak friction regime, one assumes that the friction timescale τ_w is long and the noise amplitude b_w is small. In this limit, the (z, w) system from (2) is a Hamiltonian system:

$$\frac{dz}{dt} = +\frac{\partial E}{\partial w},$$

$$\frac{dw}{dt} = -\frac{\partial E}{\partial z},$$
(15)

where the energy E(z,w) from (9) plays the role of the Hamiltonian function. In general, Hamiltonian systems are systems of the form $\frac{d}{dt}(p,q) = (\partial H/\partial q, -\partial H/\partial p)$ which are completely determined by a scalar function of position q = z and momentum p = mw. The mass *m* here is assumed constant. For our specific case, the resulting Hamiltonian is $H(p,q) = \frac{p^2}{2m} + m\Pi(q)$. Dividing everything by *m*, the Hamiltonian system can be rewritten as in (15) with a function of velocity and position E(z,w). Hamiltonian systems have many elegant and illuminating properties, as discussed further below.

Another property of Hamitonian systems is described by Liouville's theorem, which states that the phase space volume of a closed surface is preserved over time. The left panel of Figure 5 shows the evolution of a cloud of 2,000 particles initially located in the region $\left\{(z,w) | \left(\frac{z-1 \text{ km}}{0.5 \text{ km}}\right)^2 + \left(\frac{w-10 \text{ m s}^{-1}}{3 \text{ m s}^{-1}}\right)^2 \le 1\right\}$. The initial positions are described by the blue dots while the evolution after 3 min are denoted with light red dots. One can easily observe that the region under the Hamiltonian flow is deformed. The numerically computed area is close to 4.71 km ms⁻¹ and remains approximately constant over time (not shown), indicating that the area is preserved. In the second example in this figure (right panel), when the parcels reach the vicinity of the saddle point (i.e., LFC), some parcels continue to ascend, while other parcels



Figure 5. Left panel: Hamiltonian contours as in Figure 2 in the sub-domain $[-22 \text{ ms}^{-1}, 22 \text{ ms}^{-1}] \times [0 \text{ km}, 4 \text{ km}]$, the initial positions of a cloud of 2,000 particles in an ellipse with center at $(z_c = 1 \text{ km}, w_c = 10 \text{ ms}^{-1})$ and radii $z_r = 0.5 \text{ km}$ and $w_r = 3 \text{ ms}^{-1}$ (blue), and their Hamiltonian evolution after 3 minutes (light red). The right panel repeats the previous example where the cloud of particles have center at $(z_c = 1 \text{ km}, w_c = 17 \text{ ms}^{-1})$.

This version of parcel dynamics is commonly used for simple predictions of cloud properties based on adiabatic rising parcel analysis (e.g., Emanuel 1994; Grabowski and Jarecka 2015). For instance, it predicts a maximum vertical velocity of $w_{max} = (2\text{CAPE})^{1/2}$, which arises when all of the potential energy is converted to kinetic energy. These types of model prediction are discussed in more detail in section 5. These simple ideas of energy

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conversion arise because the model is a Hamiltonian system, and
 therefore the parcel trajectories preserve total energy and follow
 the contours in Fig. 2 (top panel).

4 While it is straightforward to use this Hamiltonian model for 5 a parcel that is rising upward from a starting point at the LFC, 6 it is less straightforward for considering a parcel that begins in 7 the boundary layer (i.e., the lower well) and escapes the lower well and rises into the upper well. For this latter situation, a small 8 9 amount of (stochastic) forcing is necessary to allow the parcel to escape from the lower well by moving across energy contours. 10 Furthermore, some small amount of friction is needed for the 11 parcel to eventually settle near the LNB, as illustrated earlier in 12 the numerical simulations in Fig. 3. 13

In summary, in a weak friction regime, we expect a time 14 evolution with the following behavior. A parcel initially located 15 near the bottom well will stay there for a while until some forcing 16 pushes it upward. If the forcing is large enough, the parcel may 17 reach the level of free convection, and it then can escape the lower 18 well and move upward, following a path close to the separatrix. 19 Fig. 2 shows an schematic of such a situation indicated by the 20 blue arrow (labeled for the printed version), and Fig. 3 shows one 21 particular numerical simulation with a small amount of friction 22 and stochastic forcing included. This will be a typical transition 23 from an unsaturated stable parcel to a saturated unstable one in 24 the weak friction regime. 25

4.2. Strong friction limit and bistability perspective

28 We now look for limiting equations for the case where the friction 29 is asymptotically strong. At the same time, in the original parcel model in (2), we also consider the buoyancy and stochastic forcing 30 to be of similar magnitude to the frictional damping. In this case, it 31 is the acceleration term, dw, that is relatively small; in the limit, it 32 can be set to zero to obtain the relation $wdt = -\tau_w \Pi'(z(t))dt +$ 33 $b_w \tau_w dW$. This is a relation that describes w as a function of z 34 (plus stochastic noise); if it is substituted into the first equation, 35 dz = w dt, then w can be eliminated from the system to give a 36 limiting equation of 37

$$dz = -\tau_w \Pi'(z(t))dt + b_w \tau_w dW.$$
⁽¹⁶⁾

Hence the strong-friction limit leads to a stochastic differential equation in terms of only one variable: the height z. We call (16) the strong friction limit of the system (2).

This derivation could be done using systematic asymptotic expansions, as described in more detail in Appendix A. A brief version of the scale analysis is provided next, in order to indicate the precise assumptions that are made in deriving (16).

The scaling assumptions can be summarized by the two small parameters

λ

$$=\frac{\tau_w}{[t]}, \quad \mu := \frac{[w]^2}{[\Pi]} \tag{17}$$

where [t], [w] and $[\Pi]$ are the time, velocity and potential energy reference scales of the system given by either (45) or (46) in Appendix A, near the lower or upper well, respectively. The strong friction condition reads

$$\mu = O(\lambda), \text{ as } \lambda \to 0.$$
(18)

58The condition $\lambda << 1$ is directly associated to the strong friction59condition. The condition $\mu = O(\lambda)$ says that the potential energy60or buoyancy has comparable contribution to the dynamics as the friction.

A bistability viewpoint can be taken for the strong-friction dynamics in (16) by identifying it with the small-mass limit of a particle in a potential well (e.g., Gardiner 2004). The potential

	Small parameter	
	$\lambda = \frac{\tau_w}{[t]}$	$\mu = \frac{[w]^2}{[\Pi]}$
Fig. 3	0.44	0.045
Fig. 4	0.044	0.045

Table 2. Study of parameter regime for Figures 3 and 4. The small parameters λ and μ were obtained according to the scales of the system near the bottom well as in (46).

energy $\Pi(z)$ defines the potential well here, and, as shown in Fig. 2, it is a double-well potential. As a result, the system is bistable in the sense that there are two equilibrium points, one in the lower well (boundary layer) and one in the upper well in the free troposphere (LNB), separated by a saddle point (LFC). If a stochastic forcing can "kick" the parcel high enough, it can potentially move from the lower well to the upper well. This bistability perspective of conditional instability is explored in more detail in section 6 in the context of convective initiation time.

The properties of the strong-friction model in (16) will be described further in section 5, along with comparisons with the properties of the weak-friction model. Before that, we analyze in more detail the assumptions and parameter values that lead to either the weak-friction or strong-friction regime.

4.3. Analysis of parameter regimes of the numerical results in Section *3*

We are now in a position to analyze the parameter regime of the realizations in Figures 3 and 4. Using the scales defined in equation (46) in Appendix A near the lower or the upper well, Table 2 shows the parameters λ and μ for Figures 3 and 4. The non-dimensional parameter $\lambda = \frac{\tau_w}{[t]}$ is large for Fig. 3 and small for Fig. 4. Regarding the parameter μ associated to the potential energy, it is small and comparable to λ in Fig. 4, which confirms that the parameters used in this figure is in the strong friction regime. For Fig. 3, μ is quite small, but λ is not small and this figure is in a somewhat weak friction regime.

4.4. Critical friction strength for crossover between weak- and strong-friction regimes

What is the value of the critical friction strength where the regime changes from weak- to strong-friction above LFC in the upper well? This estimation is obtained now by determining under what conditions we have either oscillatory or decaying solutions in approximate linearized models, similar to damped harmonic oscillators.

To obtain a linearized model, start from the original parcel model in (2) written in terms of z alone as

$$z''(t) + \frac{1}{\tau_w} z'(t) + \Pi'(z(t)) = 0,$$
(19)

where the stochastic forcing has been ignored. Near the level of neutral buoyancy, $z \approx \text{LNB}$, the buoyancy force can be linearly approximated as $\Pi'(z) \approx \Pi''(\text{LNB})(z - \text{LNB})$, giving

$$z''(t) + \frac{1}{\tau_w} z'(t) + \Pi''(\text{LNB})(z - \text{LNB}) = 0, \qquad (20)$$

which is the equation of a linear damped harmonic oscillator.

This linear model has well-known solution behavior which could either be oscillatory (as in the weak-friction regime) or a non-oscillatory decaying behavior (as in the strong-friction regime). The crossover between oscillatory and decaying behavior occurs for a critical relaxation $\tau_{w,crit}$ when the discriminant

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vanishes, and the critical damping parameter is

$$\tau_{w,\text{crit}} = \frac{1}{2\sqrt{\Pi''(\text{LNB})}} \approx \frac{\text{LNB} - \text{LFC}}{2\sqrt{2\text{CAPE}}}.$$
 (21)

The latter expression arises from a quadratic approximation of the potential energy $\Pi(z)$, and it characterizes the critical damping time scale in terms of the environmental thermodynamic state in particular, in terms of the LNB, LFC, and CAPE.



Figure 6. Separatrix (thin solid line), upper contours and parcel's trajectories for $b_w = 4.91 \times 10^{-1} \text{ ms}^{-1} \text{ s}^{-1/2}$, and three values of τ_w : $\tau_w = \frac{1}{2} \tau_{w,\text{cit}}$ (dot-dashed line), $\tau_w = \tau_{w,\text{crit}} = 8.94 \times 10^{-1}$ min (solid line), and $\tau_w = 3\tau_{w,\text{crit}}$ (dashed line).

To illustrate the critical damping rate and the crossover from weak- to strong-friction, Fig. 6 shows three parcel trajectories based on three values of τ_w : $\tau_w = \frac{1}{2}\tau_{w,crit}$ (dot-dashed line), $\tau_w = \tau_{w,crit} = 8.94 \times 10^{-1}$ min (solid line), and $\tau_w = 3\tau_{w,crit}$ (dashed line). The noise amplitude $b_w = 4.91 \times 10^{-1}$ ms⁻¹ s^{-1/2} has been reduced by a factor of 10 compared to the value in Fig. 3. We observe an oscillatory behavior when the relaxation timescale exceeds the critical value. We note that such critical value is based on a linear approximation of Π' around the level of neutral buoyancy in the upper well. Values of τ_w below the critical value could be considered to be in the strong friction regime.

5. Maximum vertical velocity, cloud-top height, and cloud-top time

47 In this section, we derive and compare several properties of rising 48 parcels in the weak- and strong-friction limits. The properties can 49 all be found analytically, and they can be related to environmental 50 parameters such as CAPE, LNB, etc. Stochastic forcing is not 51 utilized in this section, and the parcel dynamics is initiated 52 at the LFC, similar to common setups of weak-friction parcel 53 ascent. The main new results in this section are the analytical 54 formulas for a parcel's adiabatic ascent in the strong-friction limit. 55 Comparisons between the weak- and strong-friction limits are 56 summarized at the end of the section. 57

5.1. Vertical velocity profile

In the weak friction limit, a parcel's properties can be found by analyzing a Hamiltonian contour, which is the path of the parcel through z - w space. A Hamiltonian contour is described by the equation $E = w^2/2 + \Pi(z)$. If the parcel follows the separatrix, it has conserved energy $E_o = \Pi(\text{LFC})$ (since the parcel initially



Figure 7. Separatrix, upper contours and parcel's trajectory for the data in Fig. 3 with $b_w = 0$. Also show are the cloud-top height, maximum velocity and the corresponding location and time it takes to get there.

has zero vertical velocity at its initial location of $z \approx$ LFC). Furthermore, one could solve $E = w^2/2 + \Pi(z)$ for w to find the vertical velocity of the parcel as a function of height:

$$w = w(z) = \sqrt{2(E_o - \Pi(z))}.$$
 (22)

In this case, the maximum vertical velocity occurs at the height of the upper well, z = LNB, and it is

$$w_{\text{max}} = \sqrt{2\text{CAPE}}$$
 at $z = \text{LNB}$. (23)

These properties are summarized in Fig. 7.

To assess the accuracy of these weak-friction approximations, Fig. 7 shows the separatrix and one parcel's trajectory. The maximum vertical velocity predicted by (23) is 49.28 ms⁻¹, compared to a value of 40.18 ms^{-1} for the example parcel trajectory. The height where this maximum occurs is z = 9.38 km, compared to the approximation LNB = 10.43 km.



Figure 8. Separatrix, upper contours and parcel's trajectory for the data in Fig. 4 with $b_w = 0$. The graph of $w = -\tau_w \Pi'(z)$ is plotted in a red solid line. Also show are the cloud-top height, maximum velocity and the corresponding location and time it takes to get there.

In the strong friction limit, on the other hand, a parcel rises according to the balance relation

$$v = w(z) = -\tau_w \Pi'(z), \tag{24}$$

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which arose above (16) in the derivation of the strong-friction limiting dynamics, and stochastic forcing has been ignored here.This equation describes the parcel's vertical velocity as a function of height, with a maximum velocity

$$w_{\max} = \max_{\text{LFC} \le z \le \text{LNB}} -\tau_w \Pi'(z) \sim \frac{\tau_w \text{CAPE}}{\text{LNB} - \text{LFC}}, \quad (25)$$

where this maximum vertical velocity is achieved at a height of (LFC+LNB)/2, the average of the LFC and LNB. These properties are summarized in Fig. 8.

To assess the accuracy of these strong-friction approximations, Fig. 8 shows one parcel's trajectory on top of the profile (25). A very good agreement is seen between the two profiles. The maximum vertical velocity and the corresponding height predicted by (25) are 11.05 ms^{-1} and 5.96 km respectively. On the other hand, the values for the example trajectory shown are 11.49 ms^{-1} and 7.44 km respectively.

5.2. Cloud-top height

For interpreting the parcel model, we use the term cloud-top height for the maximum height achieved by the rising parcel.

In the weak friction limit, the cloud top height according to (22) is approximated as

$$z = \Pi^{-1}(E_o) \sim \text{LFC} + 2(\text{LNB} - \text{LFC}) = 2\text{LNB} - \text{LFC}, \quad (26)$$

where the inverse is taken for z > LFC. For the parameter values used here, the estimated cloud-top height in equation (26) is roughly near z = 15 km, and, for comparison, the parcel's maximum height in the example trajectory in Fig. 7 is 13.85 km. In the strong friction limit, on the other hand, the profile in (24)

shows us that the cloud top height can be approximated as

$$z = \text{LNB},\tag{27}$$

where the LNB here is z = LNB = 10.43 km. For comparison, for the example trajectory in Fig. 8, the maximum height is 10.63 km.

5.3. Cloud-top time

We use the term "cloud-top time" to refer to the time taken for a parcel to rise from the LFC to its maximum height.

In the weak-friction regime, a simple approximation of the cloud-top time can be found through an approximation of the Hamiltonian dynamics. In particular, if the potential energy $\Pi(z)$ can be approximated by a quadratic function, then the dynamics is

$$\frac{d^2 z}{dt^2} \approx -\Pi''(\text{LNB})(z - \text{LNB}), \qquad (28)$$

which is the equation for a simple harmonic oscillator, with oscillation period of $2\pi/[\Pi''(\text{LNB})]^{1/2}$. Under this approximation, the parcel reaches its top height at about half the period, which is

$$T(z_o \to c) \approx \frac{\pi}{\sqrt{\Pi''(\text{LNB})}} \sim \frac{\pi(\text{LNB} - \text{LFC})}{\sqrt{2\text{CAPE}}},$$
 (29)

where c is the cloud top height.

In the strong friction regime, on the other hand, we use the limiting equation (16). Here, a simple approximation of the cloud-top time can also be found if the potential energy $\Pi(z)$ is approximated by a quadratic function, in which case the dynamics is

$$dz \approx -\tau_w \Pi''(\text{LNB})(z - \text{LNB})dt + b_w \tau_w dW, \qquad (30)$$

which is an Ornstein-Uhlenbeck process. The mean trajectory of this approximated process has an analytical formula of

$$\bar{z}(t) = \text{LNB} - (\text{LNB} - z_o) \exp\left(-\tau_w \Pi''(\text{LNB})t\right).$$
(31)

Here the initial position of the parcel is located at the level of free convection. Notice that, in the current approximation, the cloud top height is identified with the level of neutral buoyancy and the mean will take an infinite amount of time to reach it; therefore, we compute the time it takes to get very close to the LNB—e.g., to 90%, 95%, or 98% of the height of the LNB, as illustrated in Fig. 9. Since the distance from LFC to the cloud top height is LNB – LFC, the time it takes to cover a factor r of the target is

$$T(\text{LFC} \to \text{LNB}) = \frac{-\log(1-r)}{\tau_w \Pi''(\text{LNB})} \sim \frac{(\text{LNB} - \text{LFC})^2}{2\tau_w \text{CAPE}}.$$
 (32)



Figure 9. Height versus time for one example trajectory with the parameter values used in Fig. 4 (dashed line) and the mean height predicted by (31). The vertical lines predict the time to reach 90%, 95% and 98% of the target.

Fig. 9 shows the height versus time for a parcel realization (dashed line) and compares it with the approximation in (31) (solid line), showing a good qualitative agreement. The vertical lines denote the time it takes to reach 90%, 95% and 98% of the target, which is the LNB. One can also observe a good agreement between the time it takes to get close to the LNB and the predictions.

5.4. Summary and comparisons

The weak- and strong-friction limits were seen above to have distinctly different properties, and one could argue that some of the properties in the strong-friction limit are actually more realistic. In particular, in the weak-friction limit (Fig. 7), the parcel rises far above the LNB, and it actually attains its maximum vertical velocity at the LNB. Such behavior is arguably unrealistic, since the LNB is typically regarded as an estimate of cloud top, and since maximum vertical velocities are typically attained below the LNB or cloud top (e.g., Takahashi and Luo 2012, 2014; Takahashi et al. 2017). These discrepancies could be regarded as due to the absence of interactions between the parcel and its environment in the traditional (weak-friction) parcel model. On the other hand, in the strong-friction limit (Fig. 8), the parcel's momentum is damped through interaction with the environment. In this limit, the parcel's behavior is arguably more realistic: the parcel attains its maximum height at the LNB and its maximum vertical velocity about half-way between the LFC and LNB. This

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suggests that the strong-friction limit provides, at the least, an interesting opposing bound on convective properties compared to the traditional weak-friction limit, and it possibly even provides a more realistic version of parcel theory. 6.

Convective initiation time and analytic approximations

Another useful quantity to estimate is the convective initiation time-i.e., the time elapsed in waiting for a parcel to rise from the boundary layer to a barrier located above in the vicinity of LFC, at which point convection will initiate. The convective initiation time should be a random quantity, since it should depend on the 12 essentially random fluctuations of boundary layer turbulence that eventually provide the "kick" necessary to lift the parcel to the 14 LFC. To model such a random process, we use the stochastic 15 version of the parcel model shown in (2).

16 In what follows in this section, an analytical approximation is 17 derived for the convective initiation time, providing yet another 18 fundamental quantity that can be derived from parcel theory. 19 Numerical simulations are also presented in order to illustrate 20 the error involved in the analytical approximation. It is also 21 noteworthy that the analytical approximation arises from using the 22 strong-friction limit of parcel theory. 23

6.1. Analytic approximation of the convective initiation time

26 The convective initiation time problem can be viewed as a 27 bistability problem. In more detail, and to define notation, note 28 that the system is endowed with a double well potential (see Fig. 2) with two stable equilibrium points, z_0 and the LNB, and 29 one unstable equilibrium point, the LFC. Initially, the parcel is 30 located in the boundary layer near z_o , and if it can rise to the 31 LFC, then it will transition to the deeper (upper) well. Therefore, 32 it is sufficient to place an absorbing barrier LFC⁺ above in the 33 vicinity of LFC and to examine the time required to reach LFC⁺ 34 (see Fig. 2 to see its location). The mean convective initiation 35 time from z_o to LFC⁺ is denoted as $T(z_o \rightarrow \text{LFC}^+)$. We can 36 then analyze $T(z_o \rightarrow \text{LFC}^+)$ from the point of view of an *escape* 37 problem from a double well potential. 38

In this situation, for the strong-friction dynamics in (16), the mean convective initiation time is approximately

$$T(z_o \to \text{LFC}^+) \sim T_o = 2\pi \sqrt{\frac{-1}{U''(\text{LFC})}} \sqrt{\frac{1}{U''(z_o)}} \exp\left(\frac{2|\text{CIN}|}{b_w^2 \tau_w}\right)$$
(33)

This formula is the classical Arrhenius formula (e.g., Gardiner 2004), and the derivation is described in Appendix B. This approximation of $T(z_o \rightarrow LFC^+)$ holds in an asymptotic limit as

$$\mu = \frac{b_w^2 \tau_w}{2|\text{CIN}|} \to 0. \tag{34}$$

Strictly speaking, it means that the relative error, which is a nondimensional quantity, satisfies

$$\frac{T - T_o}{T_o} = O(\mu), \text{ as } \mu \to 0,$$
(35)

where T depends on μ and T_o is the limiting value for $\mu \to 0$. In 56 fact, we can compute the second term in the asymptotic expansion 57 as 58

$$\frac{T - T_o}{T_o} = \left(\frac{1}{8} \frac{U^{(4)}(\text{LFC})}{(U''(\text{LFC}))^2} - \frac{1}{8} \frac{U^{(4)}(z_o)}{(U''(z_o))^2} - \frac{5}{24} \frac{(U'''(\text{LFC}))^2}{(U''(\text{LFC}))^3} + \frac{5}{24} \frac{(U'''(z_o))^2}{(U''(z_o))^3}\right) \frac{b_w^2 \tau_w^2}{2} + O\left(\mu\right)^2.$$
(36)

Table 3. Study of parameter regime for Figures 3 and 4. The small parameters λ and μ were obtained according to the scales of the system near the bottom well as in (45).

This small parameter μ is the same one that is described in (17) and Appendix A. Physically, a small value of μ could correspond to relatively strong CIN and/or relatively strong friction (small τ_w), and a consequence of this scenario is a relatively long time until convection is initiated.

The formula in (33) is potentially valuable because it describes a quantity of interest (convective initiation time) in terms of environmental parameters, as well as friction rate and turbulent intensity. In this sense, it provides yet another simple prediction, in addition to the quantities described above in section 5 (maximum vertical velocity, cloud-top height, etc.), provided by parcel models.

Numerical illustration for the strong friction limit 6.2.

To illustrate the asymptotic approximation in (33), Fig. 10 shows the (t, z(t)) trajectory for one parcel realization that escapes the bottom well, a vertical line located at the convective initiation time ("-+") and a vertical line located at the approximated convective initiation time (33). The evolution of height with respect to time is obtained using the limiting equations (16) and the parameter values used in Figures 3 and 4. That is, $\tau_w = 7.5 \text{min}, b_w = 4.91 \times$ $10^{-1} \text{ ms}^{-1} \text{ s}^{-1/2}$ for the top panel and $\tau_w = 0.75 \text{min}, b_w =$ 1.55 ms⁻¹s^{-1/2} for the bottom panel. The convective initiation time is computed using a Monte-Carlo approach with 5000 realizations. The two vertical lines are close to each other in both cases, which indicates that the approximation in (33) is accurate.

The values of λ and μ for this case are summarized in Table 3. The values are calculated using the scales near the bottom well in equation (45) in Appendix A. One can observe a reasonably small value of μ in both cases. On the other hand, the parameter λ is small only in the case of Fig. 4 (so also Fig. 10 bottom). Hence the parameter values for Fig. 10, top panel, are in the weak friction regime, while those used in Fig. 10, bottom panel, correspond to the strong friction regime for the bottom well.



Figure 10. Convective initiation time for the values in Figures 3 (top) and 4 (bottom). In each panel one realization computed with the limiting equations (16) is shown. The vertical line with plus signs indicate the mean convective initiation time statistically computed using 5000 realizations. The vertical line with '*' show the approximation used in (33).

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Figure 11. One example realization (dashed line), numerical mean convective initiation time (vertical "-+" line) and the approximated mean convective initiation time in equation (33) (vertical "-*" line). The relaxation timescale is $\tau_w = 3.75$ min. The noise coefficients are given by $b_w = 1.15$ ms⁻¹s^{-1/2} (top panel), $b_w = 0.936$ ms⁻¹s^{-1/2} (middle panel) and $b_w = 0.66$ ms⁻¹s^{-1/2} (bottom panel).

		Error	analysis		
Fig. 11	μ	T_o	$ T - T_o $	$\frac{ T-T_o }{T_o}$	$\frac{ T-T_o }{T_o\mu}$
Тор	0.97	0.19 h	1.74 min	0.15	0.16
Middle	0.65	0.32 h	1.57 min	0.08	0.13
Bottom	0.32	1.47 h	4.72 min	0.05	0.16
Fig. 10					
Тор	0.36	0.56 h	1.72 min	0.05	0.14
Bottom	0.36	5.57 h	23.62 min	0.07	0.20

Table 4. Small parameter μ (first column), relative error (second column) and its ratio (third column). The small parameter was obtained according to the scales of the system near the bottom well given by (45).

6.3. Error analysis and other parameter values

40 To analyze the error involved in the asymptotic approximation, 41 we consider a sequence of test cases for different values of the parameter b_w . In this parameter study, we consider different values 42 from those used in Fig. 10. In all cases, the relaxation timescale 43 is fixed at $\tau_w = 3.75$ min. In each panel of Fig. 11, we show one 44 example realization (dashed black line) for the particular value of 45 the parameter b_w used in that panel. In descending order, Fig. 11 46 shows the results for $b_w = 1.15 \text{ms}^{-1} \text{s}^{-1/2}$ (top panel), $b_w =$ 47 $0.936 \text{ms}^{-1} \text{s}^{-1/2}$ (middle panel) and $b_w = 0.663 \text{ms}^{-1} \text{s}^{-1/2}$ 48 (bottom panel). Such values correspond to correlation amplitudes 49 of $\sigma_w = 12.17 \text{ ms}^{-1}, 9.94 \text{ ms}^{-1}$ and 7.03 ms^{-1} respectively. In 50 addition, two vertical lines are included, one corresponding to 51 the approximated convective initiation time ('-*') and the other 52 denoting the convective initiation time computed using a Monte-53 Carlo approach with 5000 simulations. The numerical mean value 54 and the asymptotic mean value are seen to be close to each other, 55 which indicates that the asymptotic mean value is a reasonably 56 accurate approximation.

57 Table 4 shows the error in the approximation T_o of the 58 convective initiation time $T(z_o \rightarrow \text{LFC}^+)$. Both the absolute and 59 relative error are shown to clarify the behavior as μ becomes 60 smaller. In particular, there appears to be the largest error for the smallest value of μ , based on the bottom panel of Fig. 11, in comparison to the other panels that have larger μ values. This is somewhat counterintuitive since one expects smaller error as μ becomes smaller, and the explanation is that it is the relative error

μ	3.24×10^{-1}	3.24×10^{-2}	3.24×10^{-3}
Rel. error	2.6×10^{-2}	1.0×10^{-3}	0.98×10^{-4}
Eq. (36)	1.0×10^{-2}	1.0×10^{-3}	1.02×10^{-4}

Table 5. Second term in the expansion. The parameter μ (first row), relative error (second row) and the second term in the asymptotic expansion of the convective initiation time given by equation (36) (third row) are shown. The parameter values taken from the bottom panel of Fig. 11 ($b_w = 0.663 \text{ms}^{-1} \text{s}^{-1/2}$) were used to compute the corresponding values in the first column of this table. The rest of the columns take reduced values of b_w to reduce μ by factors of 10 and 100 in columns 2 and 3 respectively.

rather than the absolute error that becomes smaller as μ becomes smaller. See (35) which shows this fact, since the convective initiation time *T* itself is a growing function of μ as $\mu \rightarrow 0$. In fact, the relative error $|T - T_0|/T_0 = O(\mu)$ decreases linearly with respect to μ according to equation (35), as shown in the last column of the table.

For the sake of completeness, we include the corresponding calculations in Table 4 for the parameter values used in Fig. 10. We note that $\mu = 0.36$ for both panels, even when τ_w and b_w are different in both cases. However, the convective initiation time for the bottom panel is almost 10 times the convective initiation time of the top panel. In any case, the ratios $|T - T_o|/(T_o\mu)$ are not too far from each other.

The second term in the asymptotic expansion of the convective initiation time (36) is verified in Table 5. The small parameter μ is specified in the first row, the relative error $|T - T_o|/T_o$ is shown in the second row and the last row displays the value of the second term in the expansion, given by the right hand side of equation (36). In the first column, we take the parameter values used in the bottom panel of Fig. 11. The rest of the columns take reduced values of b_w to reduce μ by a factor of 10 with respect to the previous column. We note that the convective initiation time grows exponentially when $\mu \rightarrow 0$, and it is quite large for the second and third columns, which are included here mainly to verify the second term in the expansion. Calculating $T(z_o \rightarrow LFC^+)$ with the Monte Carlo approach is not possible in practice for these cases. Instead, we compute the exact T given by equation (51). We note that for small enough values of μ , a very good agreement between the relative error and the second term in the expansion is observed.

6.4. Distribution of escaping times



Figure 12. Probability density function of the time where the parcel reaches the absorbing barrier $z = LFC^+ = 3.43$ km (left), starting in the lower well (z = 0.5 km), averaged over an ensemble of 5000 realizations. The red solid curve denotes the exponential distribution in equation (37). Here the parameter values were taken from the bottom panel of Fig. 11 ($b_w = 0.663$ ms⁻¹s^{-1/2} and $\tau_w = 3.75$ min), which gives $T_o = 1.47$ h. The right panel repeats the graph for the barrier z = 12 km (right) and T_o is replaces by $T_1 = 2.06$ h.

Beyond the mean value, one can seek further details by analyzing the distribution of escaping times. Fig. 12 (left panel) shows a p.d.f. of the time where the parcel reaches the absorbing barrier located at $z = LFC^+ = 3.43$ km, which is 0.5 km above the level of free convection (top) and z = 12 km (bottom), starting

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in the lower well (z = 0.5 km), averaged over an ensemble of 5000 realizations. The parameter values for this p.d.f. were taken from Fig. 11 (bottom panel). The red curve shows the exponential distribution with mean T_o , given by

$$\rho(t) = \frac{1}{T_o} e^{-t/T_o} H(t),$$
(37)

where *H* is the Heaviside function. One can observe a very good agreement between the p.d.f. and the red curve, giving evidence that the distribution of escaping times is exponential with mean convective initiation time, approximated by T_o . The right panel considers the time needed for the parcel to reach a higher altitude, as an estimate of the time needed to reach cloud top. Replacing the mean T_o by the mean of the cloud top time $T_1 = 2.06$ h, we also observe an exponential distribution.

These results suggest that the convective initiation time could be modeled (e.g., as part of a convective trigger in a convective parameterization) as a random variable with an exponential distribution. The exponential distribution is a one-parameter distribution, where the one parameter can be taken to be the mean, which here could be specified using the asymptotic formula in (33).

6.5. Validation of the convective initiation time in the full system



Figure 13. Top: Convective initiation time for the values in Fig. 3 using the full system. Bottom: Convective initiation time for the values in Fig. 4 using the full system.

43 The earlier parts of this section were based on the strong-44 friction limiting dynamics in (16) as the starting point, from 45 which the asymptotic approximation in (33) was derived. Now 46 we go back one step further and compare the asymptotic 47 approximation with the full system in (2). One expects the 48 asymptotic approximation to describe the full system in (2) when 49 both λ and μ are small—i.e., when the full system is in the strong-50 friction regime. Some test cases are now described to compare 51 with the full system for some different values of λ and μ that are 52 small or not too small.

53 Fig. 13 shows a realization (t, z(t)) for the full system (2), the 54 numerically computed mean convective initiation time (line "+") 55 obtained with a Monte-Carlo approach using 5000 realizations 56 and the corresponding approximation (33) (line "-*"). The top 57 panel uses the parameter values in Fig. 3: $\mu = 0.36$ and $\lambda = 1.41$ 58 (see Table 3). The relative error $|T - T_o|/T_o$ is 3.26, which is quite 59 large. This is consistent with the fact that the full system is not in 60 the strong friction regime and as a consequence the approximation (33) is not accurate. On the other hand, the bottom panel shows the corresponding quantities for the parameter values used in Fig. 4, which are closer to the strong friction regime: $\mu = 0.36$ and $\lambda = 0.14$ (see Table 3). Even though the buoyancy term parameter

 $\mu = 0.36$ is not too small, the approximation (33) is still valid for the full system in this case. This can be corroborated with the relative error, which is 0.22 in this case.

7. Discussion

7.1. Alternative parameterizations of friction

In the text above, the friction was assumed to take the linear-in-w form of $\tau_w^{-1}w$ in order to simplify some calculations, but other choices of the friction could be used to obtain similar results. For instance, it is common to use frictional damping that is quadratic in w (e.g., Simpson and Wiggert 1969; Jakob and Siebesma 2003; Brast *et al.* 2016), which leads to a vertical velocity equation of the form

$$\frac{dw}{dt} = -\epsilon_{ent}(z)|w|w + b(z), \tag{38}$$

where $\epsilon_{ent}(z)$ is the entrainment coefficient. (Note that we write the drag as |w|w so it decelerates a parcel that is either rising or falling; it is more commonly written as w^2 in the literature in contexts where the parcel is assumed instead to be always rising.)

In the strong-friction limit, (38) would become

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$$e_{nent}|w|w = b(z), \tag{39}$$

or, upon solving for w,

$$\frac{dz}{dt} = \operatorname{sgn}[b(z)] \sqrt{\frac{|b(z)|}{\epsilon_{ent}(z)}},$$
(40)

where w = dz/dt was used to arrive at an equation for z(t) alone. Also used to arrive at (40) is the relation sgn(w) = sgn[b(z)], deduced from (39), where sgn is the sign function. The result in (40) is the quadratic-in-w analog to the linear-in-w dynamics of (16), and the main difference is the square root. For the quadratic-in-w case, similar results could be obtained for the bistability perspective, convective initiation time, etc.

7.2. Entrainment of moisture and equivalent potential temperature

Adiabatic parcel dynamics have been assumed here, where the values of q_t and θ_e are conserved, in the main results of the paper. It would be interesting in the future to investigate non-adiabatic cases with entrainment of moisture and equivalent potential temperature in which case the dynamics of the parcel's thermodynamic variables could be modeled as

$$\frac{d\theta_e}{dt} = -\frac{1}{\tau_{\theta}}(\theta_e - \theta_{e,env}(z)), \qquad (41a)$$

$$\frac{dq_t}{dt} = -\frac{1}{\tau_q}(q_t - q_{t,env}(z)), \qquad (41b)$$

where entrainment would relax the parcel's θ_e and q_t values toward their environmental values of $\theta_{e,env}$ and $q_{t,env}$, and where a constant entrainment coefficient has been written for illustrative purposes but could be replaced by a more sophisticated parameterization without causing any major change to the discussion.

Three limits are possible for (41). First, if the entrainment is weak, then the adiabatic case is obtained, as used here in the main results of the paper. Second, if the entrainment is strong, then the parcel's θ_e and q_t values are rapidly relaxed toward the environmental values of $\theta_{e,env}(z)$ and $q_{t,env}(z)$, so that $\theta_e \approx \theta_{e,env}(z)$ and $q_t \approx q_{t,env}(z)$; this case would perhaps be uninteresting since the buoyancy would be small or zero. Third, if the entrainment is O(1), then the evolution of θ_e and q_t must be retained, with entrainment, as in (41).

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1 While the third case is more complicated, the first case 2 (the adiabatic case) allows a very convenient simplification: the 3 dynamics is a gradient system. This means that the dynamics 4 can be written as dz/dt = dF/dz, where F(z) is a function of 5 z. In this case, some quantities such as convective initiation time 6 can be found analytically, as shown above. On the other hand, in the third case where entrainment is O(1) and θ_e and q_t are 7 evolving as in (41), the dynamics are a nongradient system. 8 This means that the dynamics do not take the form dz/dt =9 $\partial F/\partial z$, $d\theta_e/dt = \partial F/\partial \theta_e$, $dq_t/dt = \partial F/\partial q_t$ for some function 10 $F(z, \theta_e, q_t)$. In such a case, it is more difficult and sometimes 11 unknown how to obtain analytical formulas for quantities such 12 as convective initiation time (Bouchet and Reygner 2016; Tao 13 2018). It would be interesting to further investigate this case in 14 the future. Moreover, it would also be interesting, following the 15 ideas of transition path theory (Metzner et al. 2006), to study not 16 only the time to initiate convection but also the pathway taken in 17 the transition. 18

7.3. Effects of comprehensive thermodynamics

21 In the text above, we used simplified thermodynamics, such as the 22 linearized formulas for θ_e and θ_v in (6), since it allows analytical 23 formulas to be obtained. However, we have also considered cases 24 with comprehensive thermodynamics, and the main results of this 25 paper are essentially the same, although some of the results can 26 only be determined numerically, not analytically.

8. Conclusions

The weak- and strong-friction limits of a parcel model were investigated here as limiting cases of the vertical velocity equation

$$\frac{dw}{dt} = b(z) - \Gamma(w, z), \tag{42}$$

along with dz/dt = w, where b(z) is the buoyancy and $\Gamma(w, z)$ is the momentum drag or friction. The weak-friction limit is the traditional limit

$$\frac{dw}{dt} = b(z),\tag{43}$$

which arises when the drag $\Gamma(w, z)$ is small. In this limit, the dynamics of z and w is a Hamiltonian system. The strong-friction limit, on the other hand, is given by

$$\frac{1}{\tau_w}\frac{dz}{dt} = b(z) + f,\tag{44}$$

45 which was derived here under the assumption of small 46 acceleration dw/dt relative to friction and buoyancy. [In writing 47 (44), it was assumed that the friction has the simple form 48 $\Gamma(w,z) = \tau_w^{-1}w$; alternative forms of $\Gamma(w,z)$ were discussed 49 in section 7.1 and lead to similar equations.] In brief, the key 49 observation is that there are essentially two important time scales: 50 the frictional time scale and the buoyancy time scale.

51 As one main result, the strong-friction limit may provide a more 52 accurate estimate of cloud-top height. In particular, the strong-53 friction limit predicts that the cloud-top height will be the LNB, 54 whereas the weak-friction limit predicts that the cloud-top height 55 will be approximately *twice as high* as the LNB. While more 56 realistic estimates of cloud-top height can also be obtained by 57 including entrainment of environmental moisture, the value of the 58 strong-friction, adiabatic model is that it can be solved exactly 59 and provides analytical formulas. For example, the parcel's 60 vertical velocity is predicted to be $w(z) \approx \tau_w(g/\theta_0)\theta'_v(z)$, which is proportional to the friction time scale τ_w and the buoyancy; the parcel reaches its maximum vertical velocity in the midtroposphere and its vertical velocity decreases toward zero as it approaches the LNB.

A potential practical use is to stochastic convective parameterizations. In particular, a stochastic trigger could be formulated based on the strong-friction limit. It was seen in section 6 that the convective initiation time is approximately an exponential random variable, and its mean value is given asymptotically by the formula in (33). To determine whether convection should be initiated during a climate model time step Δt , one could draw exponential random variables for the convective initiation time and check to see whether they are less than μt —i.e., whether convection has initiated during the time step.

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A. The limiting strong friction dynamics

The goal of this appendix is to present the details in the derivation of the limiting equations for the strong friction regime. For instance, under what conditions on CAPE, CIN, and frictional dissipation rate is the strong-friction limit valid?

A.1. The scales of the system

One needs to choose the scales of the system based on the parameters involved in the equations. We note that the parcel's dynamics may exhibit different behavior below and above the level of free convection, so we identify two different sets of scales, one for each well, and given by

$$[\Pi] = |\text{CIN}|, [z] = \text{LFC} - a, [w] = b_w \sqrt{\frac{\tau_w}{2}}, [t] = \frac{[z]}{[w]}$$
(45)

for the lower well and

$$[\Pi] = \text{CAPE}, [z] = \text{LNB} - \text{LFC}, [w] = b_w \sqrt{\frac{\tau_w}{2}}, [t] = \frac{[z]}{[w]}$$
(46)

for the upper well. We note that the velocity scale is of stochastic type in the sense that it is the quasi steady state standard deviation if we were under an Ornstein-Uhlenbeck process. On the other hand, the scale for the potential energy is of deterministic type in the sense that |CIN| and CAPE are related to the depth of the bottom and upper wells and they determine the periods of the parcels if we only consider the Hamiltonian system with no relaxation or stochastic noise. The quantity $\Pi'(z)$ is non-dimensionalized with the scale $[\Pi'] = [\Pi]/[z]$. For the potential energy used in this work, $z_o = 0.5$ km, LFC = 2.85 km, LNB = 10.43 km, CIN = 152.18J kg⁻¹ and CAPE = 1214.2J kg⁻¹.

We observe that the relaxation timescale τ_w for Figures 3 and 4 differ by a factor of 10, and the corresponding coefficient b_w by a factor of $1/\sqrt{10}$ such that the stationary standard deviation coincide. As a result, the velocity scales are the same for those two parameter values. One could think of other velocity scales such as the maximum velocity given by the separatrix near each well, which would be of deterministic type. We expect that the velocity scale chosen in (45) or (46) is the smaller of the two. For the parameters used in Figures 3 and 4, the velocity scale is $[w] = 7.37 \text{ ms}^{-1}$. The length scales are given by [z] = 2.36 km and [z] = 7.57 km for the lower and upper wells, respectively.

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A.2. Nondimensional parameters and distinguished limit

To identify the nondimensional parameters μ and λ and how they they are related in the strong-friction limit, it is convenient to work with the Fokker-Planck equation rather than the stochastic differential equation itself. Let P(z, w) be the probability density function of system (2) and let us recall that $b = \frac{g}{\theta_o} \theta'_v = -\Pi'(z)$. The corresponding Fokker-Planck equation is given by

$$\frac{\partial P}{\partial t} = -w\frac{\partial P}{\partial z} + \Pi'(z)\frac{\partial P}{\partial w} + \frac{1}{\tau_w}\frac{\partial(wP)}{\partial w} + \frac{b_w^2}{2}\frac{\partial^2 P}{\partial w^2}.$$
 (47)

Using the definitions in (17) and the relation $\frac{b_w^2[t]/2}{[w]^2} = \frac{[t]}{\tau_w} = \lambda^{-1}$, the non-dimensional version of equation (47) becomes

$$\frac{\tilde{\partial}P}{\partial\tilde{t}} = -\tilde{w}\frac{\partial\tilde{P}}{\partial\tilde{z}} + \mu^{-1}\tilde{\Pi}'(\tilde{z})\frac{\partial\tilde{P}}{\partial\tilde{w}} + \lambda^{-1}\frac{\partial(\tilde{w}\tilde{P})}{\partial\tilde{w}} + \lambda^{-1}\frac{\partial^{2}\tilde{P}}{\partial\tilde{w}^{2}}.$$
(48)

This equation suggests the distinguished limit $\mu = O(\lambda)$, as presented in the main text in (18), in order to have comparable strength of the terms for friction, stochastic noise, and buoyancy (Π').

B. Asymptotic analysis for the mean convective initiation time

The topic of this appendix is a simplified formula for the average convective initiation time, in an asymptotic limit for strong friction regime. In particular, in the derivation here, the connection is described with the two small parameters λ and μ of the strong-friction limit, and the error term is presented explicitly.

To simplify notation, it is convenient to define the potential Uand the parameter D such that

$$U = \tau_w \Pi, \ b_w = \frac{\sqrt{2D}}{\tau_w}.$$
 (49)

Here U and D are in units of $m^2 s^{-1}$. In terms of U and D, the strong-friction system from(16) can then be written as

$$dz = -U'(z)dt + \sqrt{2D}dW.$$
(50)

Given the dynamics in (50), the starting point for the asymptotics is the exact formula for the mean convective initiation time (e.g., Gardiner 2004)

$$T(z_o \to \text{LFC}^+) = \frac{1}{D} \int_{z_o}^{\text{LFC}^+} \left[e^{U(z)/D} \int_{-\infty}^z e^{-U(z')/D} dz' \right] dz,$$
(51)

where LFC⁺ > LFC is the absorbing barrier located just above the level of free convection (see bottom panel of Fig. 2). Let us define the scales of the system [z], [t], and [w] as in Appendix A, and U is non-dimensionalized with the scale $[U] = \tau_w[\Pi]$. Here we note that the following relations hold:

$$\mu = \frac{D}{[U]}, \quad \frac{D}{([z]^2/[t])} = \frac{\tau_w}{[t]} = \lambda.$$
(52)

The asymptotic approximation is obtained with the use of the Laplace method. Denoting again with tildes the non-dimensional quantities, equation (51) can be re-written as

$$\tilde{T}(\tilde{z_o} \to \tilde{\text{LFC}^+})$$

$$= \frac{1}{\lambda} \int_{\tilde{z}_{o}}^{\widetilde{LFC^{+}}} e^{\mu^{-1}\tilde{U}(\tilde{z})} d\tilde{z} \int_{-\infty}^{\widetilde{LFC^{+}}} e^{-\mu^{-1}\tilde{U}(\tilde{z}')} d\tilde{z}'$$

$$- \frac{1}{\lambda} \int_{\tilde{z}_{o}}^{\widetilde{LFC^{+}}} \left[e^{\mu^{-1}\tilde{U}(\tilde{z})} \int_{\tilde{z}}^{\widetilde{LFC^{+}}} e^{-\mu^{-1}\tilde{U}(\tilde{z}')} d\tilde{z}' \right] d\tilde{z}.$$
(53)

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We note that the second term is integrated over the domain $\tilde{z_o} \leq \tilde{z} \leq \widetilde{\text{LFC}^+}, \tilde{z} \leq \tilde{z}' \leq \widetilde{\text{LFC}^+}$. It is important to choose LFC⁺ close enough to LFC so that $U(\text{LFC}^+) > U(z_o)$ (see Fig. 2). The potential \tilde{U} is increasing from $\tilde{z_o}$ to LFC, which implies $\tilde{U}(\tilde{z}) - \tilde{U}(\tilde{z}') \leq \tilde{U}(\text{LFC}^+) - \tilde{U}(\tilde{z_o}) < U(\text{LFC}) - \tilde{U}(\tilde{z_o})$, leading us to the following inequality

$$\frac{1}{\lambda} \int_{\tilde{z}_{o}}^{\widetilde{\text{LFC}^{+}}} \int_{\tilde{z}}^{\widetilde{\text{LFC}^{+}}} e^{\mu^{-1}(\tilde{U}(\tilde{z}) - \tilde{U}(\tilde{z}'))} d\tilde{z}' d\tilde{z}
\leq \underbrace{(\widetilde{\text{LFC}^{+}} - \tilde{z}_{o})^{2}}_{2\lambda} e^{\mu^{-1}(\tilde{U}(\widetilde{\text{LFC}^{+}}) - \tilde{U}(\tilde{z}_{o}))}.$$
(54)

We will later see that the leading order term is exponentially increasing when $\mu \to 0$ with exponent $e^{\mu^{-1}(\tilde{U}(\tilde{LFC})-\tilde{U}(\tilde{z}_o))}$, and so the error above can be ignored.

The other terms are estimated as follows. The integral with the negative sign in the exponent reaches its maximum at $\tilde{z}' = \tilde{z_o}$ while the other expression with the negative exponent reaches its maximum at $\tilde{z} = \widehat{\text{LFC}}$. Using Laplace's method, we get the following asymptotic expansion as $\mu \to 0$

$$\int_{-\infty}^{\widehat{\text{LFC}^+}} e^{-\mu^{-1}\tilde{U}(\tilde{z}')} d\tilde{z}' = \sqrt{\frac{2}{\tilde{U}''(\tilde{z}_o)}} \mu^{1/2} \Big[1 + \Big(\frac{5}{24} \frac{(\tilde{U}''(\tilde{z}_o))}{(\tilde{U}''(\tilde{z}_o))^2} - \frac{1}{8} \frac{\tilde{U}^{(4)}(\tilde{z}_o)}{(\tilde{U}''(\tilde{z}_o))^2} \Big) \mu + O(\mu^2), \Big]$$
(55)

and

$$\int_{\tilde{z}_{o}}^{\tilde{LFC}^{+}} e^{\mu^{-1}\tilde{U}(\tilde{z})} d\tilde{z} = \sqrt{\frac{-2}{\tilde{U}''(\widetilde{LFC})}} \mu^{1/2} \Big[1 + \left(\frac{-5}{24} \frac{(\tilde{U}''(\widetilde{LFC}))}{(\tilde{U}''(\widetilde{LFC}))^2} + \frac{1}{8} \frac{\tilde{U}^{(4)}(\widetilde{LFC})}{(\tilde{U}''(\widetilde{LFC}))^2} \right) \mu + O(\mu^2) \Big].$$
(56)

The mean convective initiation time is then asymptotically expanded as

$$\tilde{T} = \frac{\mu}{\lambda} e^{\mu^{-1}(\tilde{U}(\widetilde{\text{LFC}}) - \tilde{U}(\tilde{z}_{o}))} \sqrt{\frac{2\pi}{\tilde{U}''(\tilde{z}_{o})}} \sqrt{\frac{-2\pi}{\tilde{U}''(\widetilde{\text{LFC}})}} \left[1 + \left(\frac{-5}{24} \frac{(\tilde{U}''(\widetilde{LFC}))}{(\tilde{U}''(\widetilde{LFC}))^{2}} + \frac{1}{8} \frac{\tilde{U}^{(4)}(\widetilde{LFC})}{(\tilde{U}''(\widetilde{LFC}))^{2}} + \frac{5}{24} \frac{(\tilde{U}''(\tilde{z}_{o}))}{(\tilde{U}''(\tilde{z}_{o}))^{2}} - \frac{1}{8} \frac{\tilde{U}^{(4)}(\tilde{z}_{o})}{(\tilde{U}''(\tilde{z}_{o}))^{2}} \right) \mu + O(\mu^{2}) \right]$$
(57)

The dimensional form of this equation is presented in the main text in Section 6.

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