A Gray-Radiation Aquaplanet Moist GCM. Part II: Energy Transports in Altered Climates

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ABSTRACT

A simplified moist general circulation model is used to study changes in the meridional transport of moist static energy by the atmosphere as the water vapor content is increased. The key assumptions of the model are gray radiation, with water vapor and other constituents having no effect on radiative transfer, and mixed layer aquaplanet boundary conditions, implying that the atmospheric meridional energy transport balances the net radiation at the top of the atmosphere. These simplifications allow the authors to isolate the effect of moisture on energy transports by baroclinic eddies in a relatively simple setting.

The authors investigate the partition of moist static energy transport in the model into dry static energy and latent energy transports as water vapor concentrations are increased, by varying a constant in the Clausius–Clapeyron relation. The increase in the poleward moisture flux is rather precisely compensated by a reduction in the dry static energy flux. These results are interpreted with diffusive energy balance models (EBMs). The simplest of these is an analytic model that has the property of exact invariance of total energy flux as the moisture content is changed, but the assumptions underlying this model are not accurately satisfied by the GCM. A more complex EBM that includes expressions for the diffusivity, length scale, velocity scale, and latitude of maximum baroclinic eddy activity provides a better fit to the GCM’s behavior.

1. Introduction

In midlatitudes, the poleward transport of heat in the atmosphere dominates over the oceanic contribution (Trenberth and Caron 2001). Dry static energy and moisture fluxes contribute more or less equally to the atmospheric flux (Trenberth and Stepaniak 2003b). Since moisture fluxes play such an important role in the total poleward energy transport, there could potentially be large changes in energy fluxes and hence temperature gradients in climates with increased moisture content. This would include, for example, global warming scenarios or climates such as the Cretaceous. However, most of our understanding of the extratropical circulation is based on dry theories. The goal of this study is to improve our understanding of the effect of moisture on energy fluxes and midlatitude eddy dynamics in general.

There are several schools of thinking regarding the effect of moisture on midlatitude atmospheric circulations. On the one hand, moisture serves as an additional source of available potential energy for baroclinic eddies; therefore, it is possible that with increased moisture content baroclinic eddies may increase in strength. On the other hand, in terms of the vertically integrated heat budget of the atmosphere, poleward moisture fluxes serve to decrease temperature gradients just as do dry static energy fluxes. If baroclinic eddies are thought of as working off of these temperature gradients, with increased water vapor concentration (and increased meridional fluxes of water vapor) one might expect a decrease in strength of baroclinic eddies.

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The theme of “compensation” between different components of the poleward energy transport, the idea that the total transport is more strongly constrained than individual components, is a recurring one in climate modeling. Manabe and coauthors have studied the changes in energy fluxes under different climate model configurations in several studies. For instance, when the ocean component is removed from a coupled general circulation model (GCM) and replaced by a mixed layer surface, the total (atmosphere plus ocean) energy fluxes are nearly unchanged (Manabe et al. 1975); the atmospheric energy transport increases to compensate the loss of ocean heat flux. When mountains are removed from a similar atmosphere-only GCM, the total moist static energy fluxes also show little change, even though the partition into stationary eddy and transient eddy fluxes is very different (Manabe and Terpstra 1974). Finally, Manabe et al. (1965) study the removal of moisture from a full GCM, albeit by artificially constraining the static stability in the dry model. Again, the moist static energy fluxes are found to be relatively invariant; there is a compensating increase in dry static energy fluxes as the moisture fluxes are removed.

Stone (1978) suggests a very simple framework for understanding the relative invariance of the atmospheric moist static energy fluxes (or total atmosphere plus ocean energy fluxes) in these studies. The claim is that the total poleward energy flux is close to that obtained by assuming that the outgoing longwave radiation (OLR) is independent of latitude, given the observed pattern of absorbed solar radiation. Relatively flat OLR is seen in GCM simulations as well as observations. The claim is that the atmosphere is efficient at flattening the OLR, and it is the proximity to this limit that results in the insensitivity of the total flux.

From another perspective, the determination of poleward energy fluxes is often studied within a diffusive framework. This body of work includes the energy balance model studies of Budyko (1969) and Sellers (1969), and the diffusivity scaling theories of Stone (1972), Green (1970), Held and Larichev (1996), Haine and Marshall (1998), and Barry et al. (2002). The local diffusivity argument is evaluated in a dry quasigeostrophic model by Pavan and Held (1996). There are alternatives to the diffusive framework. For instance, baroclinic adjustment theories predict the temperature structure of the atmosphere (based on neutralizing some measure of baroclinic instability) without references to the fluxes required to give this structure. Diffusive theories in which the diffusivity is a strong function of temperature gradients predict that it is hard to change these gradients, and baroclinic adjustment theories can have the same result. To the extent that the OLR is primarily a function of temperature and the absorbed solar radiation is fixed, the total atmospheric energy transport would be hard to change as well.

Our goal is to evaluate these differing perspectives within an idealized moist GCM by varying the water vapor content of the atmosphere, and studying the changes in moist static energy fluxes and their partition into dry static energy fluxes and moisture fluxes.

An idealized moist GCM

Water has a remarkable variety of effects on the climate (Pierrehumbert 2002). We have developed a simplified moist GCM that isolates one of these, the dynamical effect of water vapor through latent heat release, in order to study the interactions of moisture with large-scale dynamics in a framework of relative simplicity.

Complete descriptions of all the physical parameterizations in our model can be found in Frierson et al. (2006, hereafter referred to as Part I). We give a brief summary of the parameterizations here. The surface boundary condition is a zonally symmetric aquaplanet with a slab mixed layer ocean of fixed heat capacity, so sea surface temperatures adjust to achieve energy balance in the time mean. There are no ocean heat fluxes in the model, dynamical or prescribed, so the atmosphere performs all the energy transport in the model.

We use a gray-radiation scheme in the model, in which the optical depths are fixed and radiative fluxes are a function of temperature alone. There are therefore no cloud- or water vapor–radiative feedbacks. This assumption allows us to study the dynamical impact of increasing or decreasing the water vapor content of the atmosphere in isolation from any radiative effects. We do not claim that the dynamical effects isolated in such a model dominate over radiative effects when, say, the climate is perturbed as in global warming simulations. But we do argue that it is very helpful to isolate the dynamical from the radiative effects in this way in order to build up an understanding of the fully interactive system. The optical depths in the gray scheme are a function of latitude and pressure, which we design to approximate the effect of water vapor in the current climate. The shortwave radiative heating approximates the annual mean net shortwave heating at the top of the atmosphere, and is all absorbed at the surface. There is no annual or diurnal cycle in the model.

Our surface fluxes are calculated from a simplified Monin–Obukhov scheme in which drag coefficients are independent of boundary layer stability provided the surface is unstable, but with reduced drag over stable surfaces. The boundary layer scheme is a standard K-
profile scheme with prognostic depth that asymptotes to the diffusivities implied by the Monin–Obukhov scaling near the surface. There is no convection scheme, only large-scale condensation when a grid box becomes saturated. There is additionally reevaporation of any falling precipitation into unsaturated regions below, making the rather extreme assumption that the column must be saturated all the way down for precipitation to reach the ground. There is no condensate in the model. These large-scale condensation-only simulations are very similar to simulations we have performed with a moist convective adjustment convection scheme (Manabe et al. 1965). We have additionally developed a simplified Betts–Miller convection scheme for use in this model (Frierson 2007); the results presented here, focused primarily on midlatitude fluxes, are insensitive to the choice of convection scheme unless otherwise stated. We utilize a spectral dynamical core with sigma coordinates, with vertical advection of water vapor by the piecewise parabolic method.

In section 2, we present results concerning the energy transports as the water vapor content is changed. In section 3, we interpret the degree of compensation of moisture fluxes by dry static energy fluxes using energy balance models, including a model with the property of exact compensation. We discuss theories for the diffusivity in these energy balance models in section 4, and conclude in section 5.

2. Dependence of energy transports on water vapor content

We vary the water vapor content of the atmosphere by changing the saturation vapor pressure constant $e^*_{0}$, which appears in the Clausius–Clapeyron relation

$$e^*(T) = e^*_0 e^{-\left(\frac{L_e}{cpR_e}T^1/T_0 - 1/T_0\right)}.$$  (1)

The latent heat $L_e$ is independent of temperature and is not varied from experiment to experiment. We find that varying the $e^*_0$ constant is a useful and simple way to vary the moisture content of the atmosphere, with higher $e^*_0$ values used as an analog for warmer climates, and lower $e^*_0$ as an analog for colder climates. In Eq. (1), we use $T_0 = 273.16$ K, and the control value of the saturation vapor pressure parameter is $e^*_0 = 610.78$ Pa. We additionally consider cases with $e^*_0 = \xi e^*_0$(control).

Experiments with $\xi = 0$ are denoted as the “dry limit”, $\xi = 10$ is the 10X case, etc., in the following. The simulations presented here are run for 1080 days, with averages calculated over every time step of the final 720 days. We have run the following cases at T85 (corresponding to 1.4° horizontal resolution) with 25 sigma levels: $\xi = 0, 0.5, 1, 2, 4,$ and 10. We have additionally run the cases $\xi = 0, 1$, and 10 at T170 (0.7° horizontal resolution) with 25 levels. The climatologies of the T170 cases are presented in detail in Part I. We use extreme values of $\xi$ to better appreciate the effect of water vapor on atmospheric dynamics over a wide parameter range. While the highest and lowest values are extreme, the moisture content in these simulations are perhaps not much more extreme than very cold and very warm climates experienced on the earth, such as the snowball earth, and the post-snowball hothouse climates, with other paleoclimate regimes and global warming scenarios intermediate.

We next study the poleward fluxes of moist static energy $m = c_p T + gz + L_d q$, dry static energy $s = c_p T + gz$, and latent energy $L_d q$. The vertically integrated flux of the moist static energy is defined as $2\pi a \cos \phi \int_0^{\infty} q dz / g$, where $a$ is the radius of the earth, $\phi$ is latitude, and the overbar denotes time and zonal mean. Figure 1 contains the vertically integrated meridional moist static energy (MSE) fluxes, dry static energy (DSE) fluxes, and latent energy fluxes as functions of latitude for $\xi = 0, 1$, and 10 at T170. The intermediate cases $\xi = 0.5, 2,$ and 4 are additionally plotted in the DSE and moisture flux plots. As we change moisture content, the moist static energy transport is nearly invariant at every latitude. The moisture transport increases greatly as we increase $\xi$, but there is a large amount of compensation by the dry static energy fluxes. A useful measure of the degree of this compensation can be obtained from the partition into sensible and latent energy fluxes at the latitude of maximum total flux. This latitude is approximately the same for all of the simulations; it is always located between 35° and 38°, as is the total (atmosphere plus ocean) poleward heat transport in observations (Trenberth and Caron 2001). The moist static energy flux and dry static energy flux at this latitude are plotted as a function of $\xi$ for all cases in Fig. 2. Table 1 lists all the fluxes at the maximum latitude for the T170 cases, and Table 2 contains the T85 cases. At the highest resolution, the amount of compensation (defined by the magnitude of the change in sensible flux divided by the change in latent flux) is almost 99% when measured from the dry limit to the control case, and decreases with increasing moisture content (to ~93% when measured from the dry limit to the 10X case). The fluxes are somewhat a function of resolution, especially for the cases with higher moisture content. We note that the compensation increases as resolution is increased. From the top-of-the-atmosphere energy budget, the invariance of the energy fluxes implies that the outgoing longwave radiation is constant as well. We discuss the
connection between these quantities, and theories for the compensation in sections 3 and 4.

The cases with moisture have a well-defined Hadley cell region out to approximately 20°, with strong equatorward transport of moisture, compensated by poleward dry static energy flux. The fluxes within this Hadley cell region only change slightly with moisture content: for instance, there is an increase of the maximum equatorward moisture flux of only \( \frac{1}{10} \times 30\% \) from the control case to the 10X case.

Poleward of the Hadley cell, the DSE fluxes decrease and shift poleward as moisture content increases. The 10X DSE fluxes in the extratropics are small (actually becoming slightly equatorward around 30°). The latitude of maximum DSE flux in the extratropics shifts significantly poleward as moisture content is increased, from 36° in the dry limit to 62° in the 10X case (Fig. 1b). The moisture flux maxima all occur at approximately the same latitude as we vary moisture, between 30° and 34° for all cases. The increase in extratropical moisture

<table>
<thead>
<tr>
<th>Simulation</th>
<th>( \phi_{\text{max}} )</th>
<th>MSE flux</th>
<th>DSE flux</th>
<th>Moisture flux</th>
<th>Compensation</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \xi = 0 )</td>
<td>35.4</td>
<td>5.61</td>
<td>5.61</td>
<td>0</td>
<td>98.9%</td>
</tr>
<tr>
<td>( \xi = 1 )</td>
<td>36.1</td>
<td>5.64</td>
<td>2.95</td>
<td>2.69</td>
<td>92.9%</td>
</tr>
<tr>
<td>( \xi = 10 )</td>
<td>36.8</td>
<td>6.03</td>
<td>0.09</td>
<td>5.94</td>
<td>92.9%</td>
</tr>
</tbody>
</table>

Fig. 1. Vertically integrated energy transports for the T170 cases (thicker lines): control case (solid), dry limit (dashed), and 10X case (dash-dot). (a) Moist static energy, (b) dry static energy, and (c) moisture. Additionally plotted in (b) and (c) are the T85 cases (thinner lines): 0.5X case (dashed), 2X case (solid), and 4X case (dash-dot). Units are PW \( (10^{15} \text{ W}) \).

Fig. 2. Vertically integrated moist static energy flux (circles) and dry static energy flux (triangles) at the latitude of maximum moist static energy flux as a function of the moisture content parameter \( \xi \). T170 simulations (filled); T85 simulations (open). Units are PW.

TABLE 1. Partition of vertically integrated moist static energy fluxes into sensible and latent components at the latitude of maximum flux \( \phi_{\text{max}} \) for the T170 simulations. Units are PW \( (10^{15} \text{ W}) \) for all simulations.
fluxes is not as rapid as the increase in \( \xi \). Between 20° and 40°, the moisture fluxes are approximately twice as large in the 10X case compared with the control case. This ratio of moisture fluxes between the 10X case and control case increases to a factor of 7 near the Pole.

The actual water vapor concentrations in the model increase more slowly than the increase in \( \xi \), especially in the Tropics, for several reasons. One important reason is that lower-tropospheric temperatures decrease with increasing moisture. The global mean tropospheric temperatures cannot change much because the insolation is unchanged, the OLR is a function of temperature only, and the average effective level of emission, where OLR = \( \sigma T^4 \), is in the midtroposphere. The vertical structure changes are primarily due to changes in the moist adiabat, forcing lower-tropospheric temperatures to decrease. Therefore, the column-integrated water only increases by a factor of 5–6 for much of the troposphere from the control case to the 10X case. In addition, the strength of the mean circulation and of the eddies decrease in the high-moisture cases. From the control case to the 10X case, the strength of the Hadley cell is reduced by a factor of 2, and the maximum eddy kinetic energy in midlatitudes decreases by over a factor of 2. These two factors combine to explain the moderate increase in moisture fluxes in midlatitudes.

Within the Hadley cell, there is an additional factor that contributes to the very modest increase of the moisture fluxes of only 30% in the 10X case. This additional effect is the moisture content at the outflow level; while negligible for the low moisture cases, the specific humidity outflow for the 10X case is half as much as the lower-layer humidities. This results from the strongly decreased lapse rates in the high-moisture cases.

Figure 3 gives the decomposition of the total flux into mean and eddy components for the T170 simulations. In observations of the current climate, the dry and latent energy fluxes, and the mean and eddy components of these fluxes, combine to create a “seamless” latitudinal profile of the MSE flux (Trenberth and Stepaniak 2003a,b). As we vary moisture content in these simulations, the partition changes drastically, but the sum still creates a seamless profile for the total flux.

While our primary focus here is on midlatitude fluxes, we briefly describe how the Hadley cell fluxes change with \( \xi \). While the total MSE transport varies smoothly in latitude and with \( \xi \), the component of the MSE transport by the mean Hadley cell varies in subtle ways. The Hadley cell actually transports energy equatorward in parts of the deep Tropics in each of these simulations, with equatorward transport over a wider area as moisture concentrations increase. This equatorward flux is balanced by strong eddy moisture fluxes in each of the moist runs, and small eddy fluxes of dry static energy in the dry case, so that the total energy transport is always poleward and smoothly varying with latitude.

A useful diagnostic when studying energy transports by the Hadley cell is the gross moist stability (Neelin and Held 1987), the amount of energy transported per unit mass transport by the circulation, that is,
with eddies carrying significant amounts of latent heat poleward. In the dry limit, in contrast, the temperatures are essentially on the same dry adiabat throughout the Tropics, implying zero static stability at the equator, and small eddy fluxes. Not surprisingly, the tropical circulation within this model, with no convection scheme, is sensitive to resolution. This dependence on resolution, and the effect of an idealized convection scheme on the Hadley circulation, gross moist stability, and tropical precipitation distribution, are studied in Frierson (2007).

3. Energy balance models

We have demonstrated in section 2 that there is a rather precise compensation between changes in latent energy fluxes and dry static energy fluxes as we vary $\xi$. We can put into context how precise this compensation is by addressing the argument of Stone (1978) that explains compensation from the flatness of the OLR. Given the net shortwave radiation at the top of the atmosphere, $S$, and the OLR, $I$, the moist static energy flux $F$ is

$$F(\phi) = 2\pi a^2 \int_0^\phi (S - I) \cos \phi^* d\phi^*,$$

where $\phi$ is latitude and $a$ is the radius of the earth. One might consider the limit in which OLR is flat, in which the atmosphere has managed to remove the temperature gradient at the effective emission level for the OLR, as the maximum sustainable atmospheric flux. This is actually not a strict upper limit in our model; we have generated solutions, with optical depths strongly decreasing poleward, with reversed temperature gradients at the emission level (but not at the surface). However, this limit is still interesting to consider. Substituting in the shortwave radiation profile used in our model, and the proper uniform OLR to ensure global energy balance (234.6 W m$^{-2}$), one obtains 7.82 PW for the maximum flux. For our idealized GCM, the fluxes are sufficiently far from this limit ($\approx$5.6 PW for $\xi = 0$ and $\xi = 1$) that this cannot be the full explanation for the insensitivity of the total flux to water vapor content.

To help interpret the compensation of energy fluxes, we introduce a one-dimensional energy balance model (EBM) in which all of the energy transport is diffusive (Sellers 1969). The form of the energy balance model is

$$0 = S - I + \tilde{D} \nabla^2 m$$

$$= S - I + \frac{\tilde{D}}{a^2 \cos(\phi)} \frac{d}{d\phi} \left[ \cos(\phi) \frac{dm}{d\phi} \right],$$

which transports energy toward the source of the convection, as in conditional instability of the second kind (CISK) theories (Charney and Eliassen 1964),
where \( \tilde{D} \) is the diffusion coefficient. It is related to the kinematic diffusivity \( D \), by \( \tilde{D} = p_s D g \), where \( p_s / g \) is the mean mass of an atmospheric column per unit area, which we set equal to \( 10^4 \) kg m\(^{-2} \) when needed. Lap-eyre and Held (2003) have obtained results that suggest lower-layer values of moist static energy are most appropriate for diffusive models of energy fluxes; therefore, we take \( m \) to be the moist static energy just above the surface, which can be written as

\[
m = c_p T_s + L_v q_s = c_p T_s + h_s L_v q_s^s
\]

\[
= c_p T_s + h_s L_v \frac{R_s e_s^s}{R_s P_s},
\]

where the subscripts \( s \) represent surface values, \( q_s^s \) is the saturation specific humidity at the surface, \( h_s \) is the surface relative humidity, and \( e_s^s \) is the saturation vapor pressure calculated using the surface temperature in Eq. (1). The dependence on \( \xi \) comes from the \( e_s^s \) expression.

a. EBM with exact compensation

The simplest energy balance model we present has the property of exact compensation: energy fluxes do not change as we vary moisture content through the parameter \( \xi \). This model uses the following assumptions in the diffusive energy balance model. First, the diffusivity \( D \) does not change as we vary moisture content. Second, the effective level of emission (the level where the temperature equals \((10 \sigma)^{1/4} \), where \( \sigma \) is the Stefan–Boltzmann constant) does not change as we vary \( \xi \). This is an excellent approximation for this GCM since the optical depths in our gray-radiation scheme are the same for all of our simulations. Third, we assume that all water vapor has condensed out at the emission level. This assumption seems reasonable since the emission level in our GCM is well above the e-folding depth for water vapor. Finally, we assume that the atmosphere has the same moist static energy at the emission level as it does at the surface. This is a reasonable starting point, since the moist isentropes in our model are close to vertical from the midlatitudes equatorward, a key result in Part I.

To prove that the solution of this EBM is independent of \( \xi \), rewrite the EBM as

\[
0 = S - \sigma T_E^4 + \tilde{D} \nabla^2 m,
\]

where \( T_E \) is the emission temperature. Using the assumptions that the atmosphere is on a moist adiabat between the surface and the emission level, and that the water vapor is negligible at the emission level, we can relate the surface moist static energy to the temperature at the emission level:

\[
c_p T_E = m - g z_E.
\]

Substituting into the EBM, we have

\[
0 = S - \sigma (m - g z_E)^4/c_p^4 + \tilde{D} \nabla^2 m.
\]

Nothing in this equation is a function of \( \xi \); therefore, there is a unique steady solution for \( m \), the flux of \( m \), and the OLR. Changing \( \xi \) only changes the partition of the fluxes into latent and sensible components. We refer to this model as EBM1.

b. Refinements to the EBM

None of the assumptions in EBM1 are exactly true in the GCM. We therefore modify the EBM to understand the GCM further. We first relax the assumption that all moisture has condensed out by the emission level, and calculate actual moist adiabats to get the emission temperature, assuming a surface relative humidity of 80% (an assumption for surface relative humidity is required, and we choose 80% as an approximation to the model simulations and observed surface humidities). We additionally assume that the diffusion coefficient is independent of latitude, and tune its value to match the maximum flux in the control case to that of the GCM. We run the EBM with the emission heights from the control run of the model as input. The emission levels in our model change little from simulation to simulation since the optical depths for radiation do not change. The emission level peaks at the equator with a value of approximately 6 km, and has a minimum at the Poles of 1.5 km. We note that the approximation

\[
z_E = 1590 - 700 \cos(\phi) + 5370 \cos^2(\phi)
\]

gives an excellent approximation to the control simulation emission level, and can be used to reproduce the results below.

The result of tuning the diffusivity to match the maximum flux in the control case to that of the GCM is \( D = 1.84 \times 10^6 \) m\(^2\) s\(^{-1}\), which produces a good fit to the flux of MSE at all latitudes. We refer to the resulting model as EBM2. The surface moist static energy gradient is close to the GCM value in the control experiment as well. One can compute the partition of the flux into moist and dry parts if one assumes that the fluxes of temperature and water vapor are individually diffusive with the same diffusion coefficient. Making this additional assumption, one finds that the ratio of moisture flux to dry static energy flux is larger than the GCM value. This can be attributed to the neglect of the moist stability, which causes the EBM surface temperatures
and hence the moisture content and moisture gradients) to be slightly too large. However, due to the large increase in complexity that would be needed to model the static stability and its spatial structure, we find this model of the control case adequate. We emphasize that adding a latitudinally constant moist stability to this model keeps the moist static energy flux the same, but does affect the partition into dry and moist components. Therefore, we primarily focus on the moist static energy fluxes in the following.

We proceed by using the same diffusion coefficient and emission level profile within the EBM while varying moisture. The maximum fluxes for these cases can be seen along with the GCM values in Fig. 5. Clearly, we have lost the precise cancellation captured by EBM1. The maximum fluxes now range from 5.26 PW for the dry limit to 7.22 PW for the 10X case. Despite its elegance, we do not believe that EBM1 captures the essence of this precise cancellation, since the assumptions made in EBM2 with regard to the difference in moist static energy between the surface and the level of emission mimic the GCM more closely.

We now examine the effective diffusivities found in the GCM, both to refine the energy balance model further and to test these values with theories for the diffusivity such as Held and Larichev (1996) and Barry et al. (2002). We define the diffusivity as the vertically averaged flux of moist static energy divided by the gradient of moist static energy at the surface. These effective diffusivities for the T170 cases are plotted in Fig. 6 for the extratropics. We have removed the deep Tropics from this plot, where the diffusive approximation is not expected to be valid. The values in the Tropics are poorly defined, but do not have a large effect on the solution in any case because the MSE flux is small there. When area averaged over the extratropics (poleward of 25°) the mean diffusivity decreases with moisture content. These mean values of the diffusivity for all cases are plotted in Fig. 7. They are somewhat sensitive to the latitudinal domain used for averaging, due to the complex spatial structures. We note, however, that the value for the control case (1.87 × 10^6 m^2 s^-1) is very similar to the value that works best in EBM2. The comparison with the T85 cases in Fig. 7 demonstrates the relative insensitivity to resolution of the inferred diffusivity.
entropy, and confirms the gradual decrease of diffusivity as moisture is increased.

We next investigate the sensitivity of EBM2 to the diffusivity, first by calculating the diffusivities required to reproduce the maximum flux in the GCM simulations. Using the T170 simulations only, we find that the values required by the EBM in the 10X case and the dry limit are very close to the actual GCM values. When the control emission height is used for all cases, the required diffusivities are 2.05 \times 10^6 m^2 s^{-1} for the dry limit, and 1.10 \times 10^6 m^2 s^{-1} for the 10X case. These required diffusivities for the EBM as a function of \( \xi \) are also plotted in Fig. 7. The agreement between required EBM diffusivity and the GCM values suggests that a theory for the change in diffusivity is the only remaining component needed to explain the compensation seen in our GCM. Changes in static stability, emission level, and the structure of the diffusivity are secondary to changes in the mean diffusivity in explaining the behavior of the GCM.

To get an idea of the sensitivity of the fluxes to the diffusivity within EBM2, we run this model over a wide range of diffusivities for the dry limit, control case, and 10X case. The maximum moist static energy flux as a function of diffusivity is plotted in Fig. 8. Each point on this plot represents one steady state of the EBM. When the diffusivity is small, the fluxes go to zero and the model is in radiative equilibrium. In the other extreme limit, the surface temperature and moist static energy have become homogenized. This corresponds to a reversed OLR gradient due to the emission height structure with latitude. The flux asymptotes to a smaller value for the 10X case due to the reduced temperature lapse rate up to the emission level, creating a smaller OLR reversal, and smaller fluxes. Provided the diffusivity is not very small, the latitudinal structure of the fluxes is very similar over this wide range of diffusivities. The maximum flux always occurs within 2° of 36° provided the diffusivities are greater than 9 \times 10^5 m^2 s^{-1}.

It is clear from Fig. 8 that the maximum fluxes are quite sensitive to changes in diffusivity at their current state. In fact, each is at approximately its most sensitive point in the domain of diffusivities. The 10X case is most sensitive to diffusivity for small values of diffusivity, due to the strong positive feedback of moisture on surface moist static energy gradients as surface temperature gradients increase. Our conclusion from this plot is that the change in diffusivity from case to case is important for the observed invariance of fluxes. While it is certainly possible that the total flux in this system is somehow constrained to remain nearly unchanged for some other reason, and that the effective eddy diffusivity then adjusts to satisfy this constraint, we do not have a candidate for this constraint and, therefore, continue by examining possible theories for the diffusivity.

4. Theory for diffusivity

Eddy diffusivity theories are based on the principle that perturbations in the quantity being mixed can be written as a mixing length times the mean gradient of the quantity, in this case, that

\[
|m'| = -L_{\text{mix}} \frac{1}{a} \frac{\partial m}{\partial \phi}.
\]

Then the eddy diffusivity is calculated as the following:

\[
D = kL_{\text{mix}}|v'|,
\]

where \( k \) is a correlation coefficient between \( v' \) and \( m \), and \( |v'| \) is the rms eddy velocity. In the past, theories have been developed using length scales including the Rhines scale (Held and Larichev 1996; Barry et al. 2002), the Rossby radius (Stone 1972), the scale with maximum growth in the Charney problem (Branscome 1983; Stone and Yao 1990), and the width of the baroclinic zone (Green 1970; Haine and Marshall 1998). These theories have used scales for the velocity fluctuation including the mean zonal wind (Stone 1972; Haine and Marshall 1998), scales based on equipartition of kinetic and available potential energy (Green 1970), or scales based on entropy production and the kinetic energy cycle (Barry et al. 2002).

In Part I, we found that the length scales of eddies in the GCM, measured by the spectrum of the vertically averaged variance of the meridional velocity, are remarkably constant, both with latitude (outside of the
Tropics), and with changes in moisture, despite the large changes in dry stability and the radius of deformation. One is tempted to view this fixed eddy scale as determined by the fixed geometry, but one can change this length scale, for example, by changing the rotation rate or by changing the baroclinicity, holding $\xi$ fixed.

The dynamical interpretation offered in Part I is that the length scale is the Rhines scale at the latitude of maximum eddy kinetic energy. We define this scale to be $L_R = L |v'|/\beta$. The latitude of maximum eddy energy moves poleward as $\xi$ increases, and the resulting decrease in $\beta$ at this latitude is essential in order for this theory to fit the GCM data, with a length scale that changes very little as $\xi$ increases. From this perspective, there is nothing fundamental about the insensitivity of the eddy length scale to $\xi$.

We first investigate whether this length scale times $|v'|$ at the latitude of maximum eddy kinetic energy (EKE) gives an adequate description of the changes in diffusivity. We plot these predicted diffusivities ($kL_R|v'| = k|v'|^{3/2} \beta^{-1/2}$) along with the average GCM diffusivities in Fig. 9. A correlation coefficient $k = 0.32$ is chosen to match the T170 control case diffusivity, and this then is used for all cases. The diffusivities agree well with the GCM for high-moisture cases, but diverge slightly at low-moisture content.

To complete this expression for the diffusivity (and hence the temperature profile and fluxes from the EBM) one needs a theory for the latitude of maximum EKE, and the RMS velocity at that latitude. In Part I, we show that the static stability can for some purposes be thought of as near neutral in terms of moist stability from the midlatitudes equatorward. However, there is some moist stability in the midlatitudes that increases as moisture is added. Further, the atmosphere is very stable in the polar regions. Lacking a simple unified theory for this behavior, and consistent with the level of complexity of the EBM as presented so far, we investigate whether a useful expression for the diffusivity can be obtained without considering how to determine the moist stability and how this moist stability affects the diffusivity.

The latitude of maximum eddy kinetic energy shifts significantly poleward as the moisture content is increased. The Eady growth rate, $\sqrt{fU/\alpha z/N}$, has been successfully employed to locate the latitude of midlatitude storm tracks (Hoskins and Valdes 1990). This depends on the stability, but we simply ignore this dependence here and assume that the structure in the meridional temperature gradient is dominant, estimating the position of maximum kinetic energy by locating the maximum in the temperature gradient at 630 hPa. These quantities are plotted in Fig. 10. This simple method captures the latitude of maximum EKE quite well. The predicted latitudes are plotted in Fig. 11.

Finally, we need an expression for $|v'|$. Stone (1972) assumes

$$|v'| = \frac{1}{f} \frac{\partial T}{\partial y},$$

which is equivalent to assuming equipartition between eddy kinetic energy and the mean available potential energy within one radius of deformation, but the end result [Eq. (12)] has no explicit dependence on static stability. Since our results are fit far better by assuming
that the eddy scale is the Rhines scale (at the latitude of maximum EKE) than with a dry or moist Rossby radius, use of this expression would appear to be inconsistent. However, Schneider (2004) shows in an idealized dry model that the static stability adjusts to keep the Rossby radius proportional to the Rhines scale, preventing a significant inverse cascade. In our moist model, in which we also do not see a significant inverse cascade, we speculate that there may be an effective moist stability that allows use of this same equipartition argument, although we do not know how to estimate this effective stability independently. If this were true in our model, it suggests that a Rossby radius could be used as a measure of eddy length scales, if we knew the proper static stability to choose.

While our justification is not very solid, we have found no other simple scaling argument that works as well. The expression Eq. (12), using the temperature gradient at 630 hPa at the latitudes of maximum EKE in the GCM (and \( f \) at the latitude of maximum EKE in the GCM as well), is compared to the vertical mean GCM \( |\nu'| \) in Fig. 12.

We next run the EBM predicting the latitude of maximum EKE, the RMS meridional velocity, the length scale, and the diffusivity. These are predicted at each time step, and the model is run until converged. The equations for this model (EBM3) are made fully explicit in the appendix, where we also describe the tuning process. The results for the fluxes in EBM3 are plotted in Fig. 13. Given our level of understanding of closures for moist eddies, this level of agreement is encouraging. Further, the movement of the latitude of maximum EKE is well predicted by this model, although somewhat exaggerated: these are plotted in Fig. 11. The predicted RMS velocities can be found in Fig. 12, and the diffusivities in Fig. 9.

5. Conclusions

We have studied the meridional fluxes of moist static energy as moisture content is increased within an idealized GCM. The moisture fluxes increase with moisture as expected; however, there is an accompanying decrease in the dry static energy flux, leaving the total moist static energy flux nearly unchanged, both in
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APPENDIX

Description of EBM3

The full energy balance model EBM3 consists of the diffusive energy balance equation, combined with a
theory for the diffusivity. The full equation in steady state is
\[
S - I + \frac{\tilde{D}}{a^2 \cos(\phi) d(\phi)} \left( \frac{\cos(\phi) \frac{dm}{d\phi}}{d(\phi)} \right) = 0, \tag{A1}
\]
where \( m \) is the surface moist static energy, \( S \) is the solar heating, \( I \) is the longwave cooling, \( \tilde{D} \) is the diffusivity, \( a \) is the earth’s radius, and \( \phi \) is the latitude. The solar heating \( S \) is specified to be the same as in the GCM. The longwave cooling \( I \) is calculated from the temperature at the emission level, that is,
\[
I = \sigma T(z_E)^4, \tag{A2}
\]
with the emission level \( z_E \) given by the approximation in Eq. (9). The temperature at the emission level is calculated from the surface moist static energy assuming a surface relative humidity of 80%, and moist adiabatic temperature structure throughout the troposphere. The moist adiabatic approximation is used in all aspects of the energy balance model.

The diffusivity \( \tilde{D} \) is calculated as proportional to a length scale and an eddy velocity scale with a constant of proportionality (discussed later):
\[
\tilde{D} = \frac{k p_s L |v'|}{g}. \tag{A3}
\]
The length scale \( L \) is the Rhines scale at latitude of maximum eddy kinetic energy,
\[
L = L_0 \left[ |v'| / \beta (\phi_0) \right]^{1/2}, \tag{A4}
\]
where \( \phi_0 \) is the latitude of maximum eddy kinetic energy and \( \beta \) is the meridional gradient of the Coriolis parameter. The velocity scale is proportional to the temperature gradient at 4 km divided by the Coriolis parameter, and at the latitude of maximum eddy kinetic energy:
\[
|v'| = \frac{v_0}{f(\phi_0)} \frac{\partial T_{4000}}{\partial y}. \tag{A5}
\]
Combining expressions (A3), (A4), and (A5), we obtain the full expression for the diffusivity
\[
\tilde{D} = D_0 \left[ f(\phi_0) \right]^{-3/2} \left[ \beta (\phi_0) \right]^{-1/2} \left( \frac{\partial T_{4000}}{\partial y} \right)^{3/2}, \tag{A6}
\]
where all relevant constants have been gathered into the constant \( D_0 \). The temperatures at 4 km are calculated from the surface moist static energy by assuming moist adiabatic ascent. Finally, we obtain the latitude of maximum eddy kinetic energy by locating the maximum temperature gradient at 4 km.

We run the model with 1000 grid points equally spaced in latitude, and integrate the equations in time until a steady state is reached. The diffusivity coefficient \( D_0 \) is calculated by tuning the flux in the control case to match the GCM value.

REFERENCES


