

Reynolds Averaged Radiative Transfer Model



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Declaration

I confirm that this is my own work and the use of all material from other sources has been properly and fully acknowledged.

Signed

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Abstract

In order for accurate predictions of climate change, it is essential that radiative transfer processes within the atmosphere are represented accurately in GCMs. The Radiative Transfer component of climate models is one of the 'rate-limiting' areas in the computation (Natraja *et al.*, 2005). While several techniques have been proposed to speed up radiative transfer calculations, they all suffer from accuracy considerations (Natraja *et al.*, 2005). A review of current techniques used within radiative transfer modelling is given followed by the development of a new, Reynolds averaged radiative transfer model. Starting with a simple case, reducing the radiative transfer equation (RTE) to exclude scattering and emission. The fluctuating terms within the RTE are Reynolds decomposed and a system of linear, homogeneous differential equations is solved involving a closure problem, which requires parameterization. Real atmospheric absorption spectra are used and an accuracy of 10^{-6} Wm^{-2} is achieved for 100 bands in an equal weighted band model, where each band has the same number of extinction coefficients within it. The models developed have difficulty in representing a large range of absorption coefficient. The next step in the development and testing of a Reynolds averaged radiative transfer model would be to include radiation travelling in different directions and to also include emission along the path, therefore having to include the Planck function. Then we would like to account for scattering. We would also like to account for an inhomogeneous path, where pressure and temperature change.

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Chapter 1

gases and clouds. These processes have a strong impact on the surface energy balance (Ritter & Geleyn, 1992).

As in many areas of numerical modelling, especially in GCMs, it is desirable for the mathematics within the model to be calculated quickly, whilst describing as accurately as possible the interaction of radiative processes (Ritter & Geleyn, 1992) (Dessler *et al.*, 1996) as even a change of 1% in radiation calculations are significant for climate (Turner *et al.*, 2004).

At present, large computational resources are required for the calculation of radiative transfer in the atmosphere for weather and climate prediction (Fomin, 2004). The radiative transfer component of climate models is one of the 'rate-limiting' areas in the computation (Natraja *et al.*, 2005). While several techniques, which will be discussed later, have been proposed to speed up radiative transfer calculations, they all suffer from accuracy considerations (Natraja *et al.*, 2005). The need to improve the computational efficiency of the numerical models causes the problem of gaining the right balance between efficiency and accuracy (Stephens, 1984). Radiative transfer schemes will have to compromise between these two criteria (Ritter & Geleyn, 1992).

As full treatment of the radiative transfer code in GCMs, incorporating all known physics is computationally expensive, parameterization is required (Turner *et al.*, 2004). The cost effectiveness of a parameterization scheme is subject to many influences. First there is the solution method to the radiative transfer equation (RTE) and the approximation associated with it. Second there is the number of intervals used to resolve the spectrum. Further economy can be achieved in many models by considering only those optical constituents of the atmosphere that are important in a particular spectral domain. Often some atmospheric constituents in certain spectral intervals are neglected where their impact is well below that of other constituents (Ritter & Geleyn, 1992).

As an example of the importance in obtaining code accurate enough to model these atmospheric interactions, there are large differences among recent GCM simulations for prescribed changes in stratospheric water vapour (stratospheric water vapour being an important contributor to the observed stratospheric cool-

ing); this points to problems with the current GCM treatment of the absorption and emission by stratospheric water vapour (Oinas *et al.*, 2001).

Even considering the current computational efficiency of radiation calculations in numerical weather prediction models and GCMs, these models still require substantial computational time relative to all other physical and dynamical calculations. As a result, radiation subroutines in these models are usually not called as often as would be desired, even being called less often than the time scale on which clouds evolve in the models. High spectral accuracy is therefore

transfer equation and applying Reynolds averaging to the variables which vary with wavelength. The main aims will be:

- Determine the accuracy of a simplified situation, no scattering or emission and with radiation travelling in only one direction.

- Investigate how the standard deviation, of the absorption coefficients affects the accuracy of the scheme

- Investigate the number of quadrature points required to achieve desired accuracy

- Investigate if bands are still required

 - Total number of bands

 - How best to define the bands

- Add emission, Planck function into equations

- A diffusivity factor, for radiation travelling in different directions

Chapter two will give an overview of the science background required to fully understand the processes within the model. Chapter three gives a summary of current techniques used to treat the complicated absorption spectrum. Chapter four gives the derivation of the equations used within the model and a description of how the model works. Chapter five presents some results from the model and chapter six discusses these results.

Chapter 2

Scientific Background

The radiative transfer process is essentially the process of interactions between matter and a radiation field (Fu & Liou, 1992). Thermal Infra Red (IR) radiation is the most important factor when considering the distribution of heat within the atmosphere. IR radiation is emitted from the Earth and the atmosphere.

temperature of the Earth's atmosphere. Hence IR radiation and the amounts absorbed and re-emitted within the atmosphere play a major role in quantifying future climate change (Liou, 1992).

2.1 Emission

The Earth and the atmosphere are continuously emitting IR radiation which may be radiated out into space or may be absorbed by other parts of the atmosphere. The emission of an object is directly related to the temperature of that object. An object of temperature T will emit over all possible wavelengths. The relationship between, temperature, wavelength and the amount of radiation emitted is defined by the Planck Function (2.1) (Petty, 2004):

$$B(\lambda, T) = \frac{2hc^2}{5(e^{hc/k_B T} - 1)}; \quad (2.1)$$

where $c=2.988 \times 10^8 \text{ m s}^{-1}$, (the speed of light), $h=6.626 \times 10^{-34} \text{ J s}$ represents Planck's constant and $k_B=1.381 \times 10^{-23} \text{ J K}^{-1}$ represents the Boltzmann's constant.

Figure (2.1) shows the shape of the Planck Function for several atmospheric temperatures. The curves describe the maximum amount of thermal radiation a body could emit, thus describing a 'black body' (one which absorbs and emits all radiation incident upon it) (Petty, 2004).

For any temperature there will be a specific wavelength at which there occurs a maximum in the amount of radiation emitted. This wavelength is defined by Wien's Displacement Law (2.2) (Petty, 2004):

$$\lambda_{max} = \frac{C}{T}; \quad (2.2)$$

where λ_{max} is the wavelength at which the peak in emission occurs and $C=2897 \text{ } \mu\text{m K}$.



Figure 2.1: The Planck Function B_λ for Blackbodies at typical atmospheric temperatures (Petty, 2004).

Equation (2.2) shows that the warmer the object, the shorter the wavelength at which the peak in emission occurs. If Planck's Function is integrated over all wavelengths we obtain Stefan-Boltzmann Law (2.4), which quantifies the total amount of radiation which can be emitted from a perfect black body (the blackbody flux F_{BB}) (Petty, 2004):

$$F_{BB}(T) = \int_0^\infty B_\lambda(T) d\lambda; \quad (2.3)$$

$$F_{BB}(T) = \sigma T^4; \quad (2.4)$$

where $\sigma = 5.67 \times 10^{-8} \text{ Wm}^{-2}\text{K}^{-4}$.

2.2 Absorption and Scattering

We can define the extinction coefficient κ_e (2.5) as the sum of the extinction due to absorption κ_a and the extinction due to scattering κ_s (Petty, 2004).

$$\kappa_e = \kappa_a + \kappa_s \quad (2.5)$$

We are now able to define the single scatter albedo ω (2.6), which defines the relative importance of scattering and absorption. ω ranges from 0 for purely absorbing mediums to 1 for purely scattering material (Petty, 2004):

$$\omega = \frac{\kappa_s}{\kappa_e} = \frac{\kappa_s}{\kappa_a + \kappa_s}$$

2.2 Absorption and Scattering

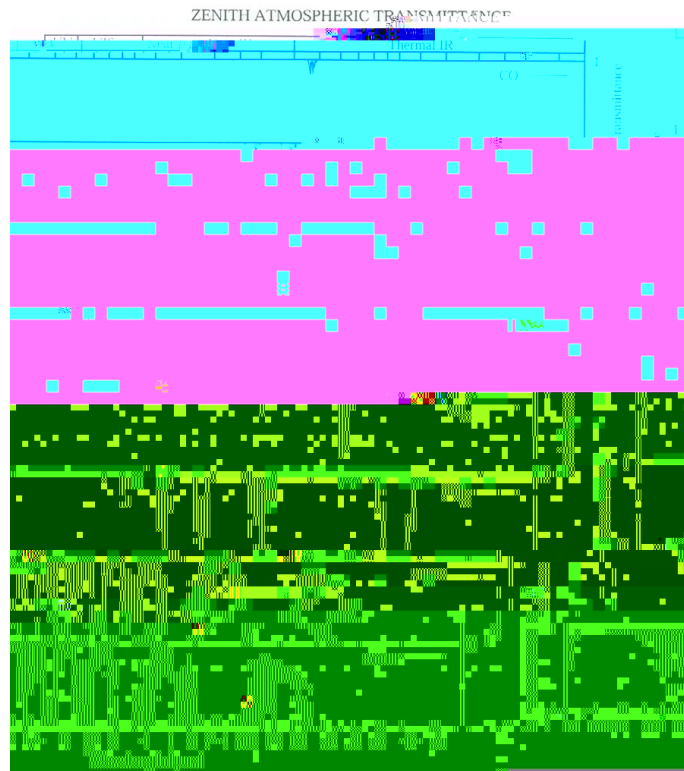


Figure 2.2: The zenith transmittance of a cloud and aerosol free atmosphere for a mid latitude summertime. The bottom panel depicts the combined effect of all the constituents above (Petty, 2004).

The most important absorbers are carbon dioxide (CO



Figure 2.3: The zenith transmittance of the atmosphere due to water vapour in the thermal IR (Petty, 2004).

The x-axis is measured in wavenumber ν , which is simply the reciprocal of the wavelength (2.7) and therefore its units are in cm^{-1} .

$$\nu = \frac{1}{\lambda}; \quad (2.7)$$

The absorption spectrum of water vapour exhibits great complexity and if one were to zoom in on figure (2.3), it would become apparent that the complexity exists on wave length scales up to 10^{-4} cm^{-1} . At wavelengths greater than 25 μm the atmosphere is more or less opaque with regards to H_2O absorption (Liou, 1992).

2.3 Radiative Transfer Equation

We are interested in the amount of radiation travelling through a part of the atmosphere. As the electromagnetic (EM) radiation transports energy, we quantify the amount of radiation transferred with power, using the Watt, W (the amount of energy measured in Joules, J, per unit time, s) (Petty, 2004). It is common to refer to the amount of radiation received in terms of its flux density F , the rate of energy transfer per unit area Wm^{-2} . From here on in flux shall be used to refer to flux density. The flux is the rate at which the radiation passes through a surface including all wavelengths between a specified range (Petty, 2004).

2.3 Radiative Transfer Equation

We shall consider how EM radiation is affected when it travels along a path of homogeneous medium and is able to be absorbed by the medium. The intensity or radiance tells us about the strength and direction of the various sources which contribute to the flux.

The flux passing through the surface will be the integral of intensity over all the possible directions from which the radiation is incident. As only one direction is normal to the surface, all contributions from other directions must be weighted by the cosine of the incident angle relative to the normal. So in spherical polar coordinates the relationship between flux and intensity can be written (Petty, 2004),

$$F = \int_0^{2\pi} \int_0^{\pi/2} I(\theta, \phi) \cos \theta \sin \theta d\theta d\phi; \quad (2.8)$$

where $I(\theta, \phi)$ is the irradiance over all possible incident angles. We shall now consider the absorption of a monochromatic EM wave propagating through a homogeneous atmosphere. The most simple governing equation for this occurs where the wave is travelling in one direction only and is attenuated only by absorption (Petty, 2004),

$$dI(z) = -\alpha I(z) dz; \quad (2.9)$$

where α is the absorption coefficient, which depends on the physical medium and the wavelength of the radiation, $I(z)$ is the intensity of the transmitted radiation at some distance, z , and I_0 is the intensity of the radiation at $z=0$. We can rearrange and integrate (2.9) over the range of intensities and the distance the radiation travels to obtain (Petty, 2004):

$$I = I_0 \exp(-\alpha z):$$

2.3 Radiative Transfer Equation

atmosphere. The beam of radiation is not only attenuated by absorption, but also by scattering due to interactions with particles. We can now use our previously defined extinction coefficient in place of the absorption coefficient to account for scattering. We consider our beam of radiation over a finite path where extinction coefficient varies with location. We replace the height, z , by s to represent a distance in any direction, represented by $\hat{g}$. (2.4).

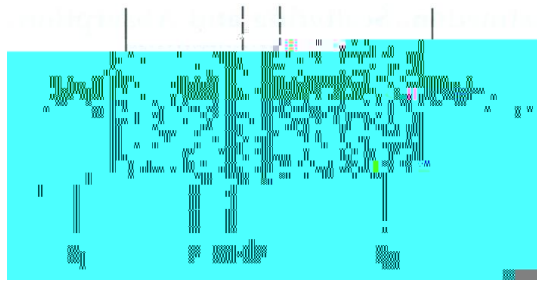


Figure 2.4: Depletion of radiation over an infinitesimal path ds (Petty, 2004)

If ds is considered to be infinitesimal, then the extinction coefficient ϵ will be constant within the interval. The radiation will then change by an infinitesimal amount dI and we can re-write (2.9) to include loss due to scattering, however no gain from scattering:

$$dI(s) = -I(s)\epsilon(s)ds: \quad (2.11)$$

We can consider the amount of radiation attenuated due to absorption only by writing (2.12) (Petty, 2004)

$$dI_{abs} = -I_a ds: \quad (2.12)$$

Neglecting scattering ($\epsilon_a = \epsilon$), we know from Kirchoff's law discussed earlier that the absorptivity of any material is equal to the emissivity of that material. We can therefore define the amount of radiation that the thin layer of air will

2.3 Radiative Transfer Equation

emit by (2.13) (Petty, 2004):

$$dI_{emit} = B(T) a ds; \quad (2.13)$$

enabling us to define Schwarzschild's Equation (2.14), the net change in radiation over the infinitesimal path in a non scattering medium (Petty, 2004):

$$dI = dI_{abs} + dI_{emit} = -a(B - I)ds; \quad (2.14)$$

We are now ready to include scattering into our radiative transfer equation. Now we have

$$dI_{ext} = S_{ext} I ds; \quad (2.15)$$

We must consider one more process to complete the RTE, in considering a new source term which includes the contributions to the beam as a result of scattering from other directions. We shall call this new source term, dI_{scat} . This term will be proportional to the scattering coefficient s . We can define the direction of travel we are interested in $\hat{\Omega}$ and we can state that any radiation from any direction $\hat{\Omega}'$ is emitted by $\hat{\Omega}'$.

We can divide through by the change in optical thickness d and can then write the complete form of the RTE (2.18) (Petty, 2004)

$$\frac{dI(\mu)}{d} = -I(\mu) + (1 - \epsilon)B + \frac{\epsilon}{4} \int_{-1}^1 p(\mu'; \mu) I(\mu') d\mu' \quad (2.18)$$

2.4 Reynolds Decomposition

Figure (2.3) of the absorption spectrum for water vapour in the thermal IR is very reminiscent of a turbulence spectrum observed in many other areas of the atmosphere, such as the variation in surface wind speed over a certain amount of time. The transmittance due to water vapour appears to vary randomly across wavelengths, however the ability to find a mean value suggests that the spectrum is not random.

We can average the transmittance spectrum over a certain, number of intervals, averaging out the positive and negative fluctuations about the mean. The mean transmittance in one interval is denoted, $\bar{T}$. At any one particular wavelength, we can subtract $\bar{T}$ from the actual transmittance, T , which will result in the fluctuating part, T' . This is represented schematically in figure (2.5) which



Figure 2.5: The deviation, T' of the actual, transmittance, T , from the local mean, $\bar{T}$

2.5 Reynolds Averaging Rules

If we let A and B be two variables that are dependent on the wavelength. We can show some rules using integration.

2.5.1 Rule One

The average of a sum (Stull, 1989):

$$\overline{A + B} = \frac{1}{N} \int_0^Z (A + B) dZ \quad (2.20)$$

$$= \frac{1}{N} \int_0^Z A dZ + \frac{1}{N} \int_0^Z B dZ \quad (2.21)$$

$$= \frac{1}{N} \int_0^Z A dZ + \frac{1}{N} \int_0^Z B dZ \quad (2.22)$$

$$= \bar{A} + \bar{B} \quad (2.23)$$

2.5.2 Rule Two

An average value acts as a constant, when averaged a second time over the same wavelength interval (Stull, 1989).

$$\frac{1}{N} \sum_{n=0}^N A_n = \bar{A}$$

2.5 Reynolds Averaging Rules

The only way the above can be true is if,

$$\overline{A'} = 0: \quad (2.33)$$

This makes sense as the sum of the positive deviations from the mean must equal the sum of the negative deviations from the mean.

2.5.5 Rule Five

Chapter 3

Literature Review

As already seen, atmospheric gases have absorption coefficients that vary rapidly as a function of wavelength or wave number ν , often changing by several orders of magnitude across the electromagnetic spectrum (Petty, 2004). This provides a great problem when the absorption spectra are modelled in GCMs, as there can be $O(10^5)$ individual absorption lines within the spectrum. If we were to model each individual line, large amounts of computer power would be required to calculate the fluxes in the atmosphere (Ellingson *et al.*, 1991). Here we shall discuss some of the techniques developed to overcome this problem and reduce the time taken for the models to run, increasing the computational efficiency, whilst attempting to maintain a suitable degree of accuracy.

3.1 Line by Line Calculations

Absorption occurring at the smallest scale, that of the line, is described by the Lorentz line absorption profile. The most straightforward and accurate but most computationally expensive way to perform radiative transfer calculations is to divide the full spectrum into monochromatic (one wavelength) intervals, or sufficiently small so as to be treated as monochromatic (10^{-4} to 10^{-2} cm^{-1}) (Ellingson *et al.*, 1991). The relative contributions of all relevant absorption lines (all lines whose wings contribute to important absorption at a particular wavelength)

3.1 Line by Line Calculations

are summed to the absorption coefficient, κ_a , for the wave number in question (Petty, 2004). Integrating over all intervals in the spectrum then provides fluxes and heating rates (Ellingson *et al.*, 1991). This however is not a simple task. The absorption spectrum depends on many things, including the location, strengths and shapes of the spectral lines (Ellingson *et al.*, 1991).

A LBL calculation of the average transmittance of a spectral interval ($\nu_1 - \nu_2$) over a finite mass path, u , can be calculated by considering (Petty, 2004):

$$T(u) = \frac{1}{\nu_2 - \nu_1} \int_{\nu_1}^{\nu_2} \exp[-k(\nu)u] d\nu; \quad (3.1)$$

which can be approximated as a sum:

$$T(u) = \sum_{i=1}^N w_i \exp[-k(\nu_i)u]; \quad (3.2)$$

where N is the number of frequencies, ν_i , where $k(\nu)$ is evaluated and w_i are weighting coefficients depending on the quadrature method used.

Because the technique involves summing the contributions from each spectral line, it is referred to as the Line-By-Line technique (LBL) (Ellingson *et al.*, 1991). If this approach was to be used to calculate heating rates in the atmosphere, the monochromatic calculation would need to be repeated for a large number of wavenumbers as well as at a number of altitudes (Petty, 2004). This means that radiative heating calculations potentially require millions of calculations to obtain a suitable accuracy (Pawlak *et al.*, 2004). Where field measurements are unavailable, LBL techniques provide the best benchmark for analysis of other numerical methods (Ellingson *et al.*, 1991; Petty, 2004). This means

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3.2 Band Models

Due to the difficulty and computational demands of treating a single Lorenz line individually, the idea of the band model was developed, which enables certain characteristics of the band to be determined using simple statistical relationships, such as, averaging of the absorption properties (Petty, 2004). An analytic approximation for the transmittance is then derived. (Stephens, 1984) suggests that the most widely used in atmospheric flux calculations is that of Goody (1952) where the absorbing lines are assumed to be randomly distributed.

3.3 k-distribution Method

3.3.1 k-distribution Method Theory

This method was first discussed by Ambartsumian (1936). It aims to perform the radiative calculations with larger discretisations by replacing the complex integration over wavenumber with an equivalent integration over a much smoother function (Petty, 2004) and involves grouping spectral intervals according to absorption coefficient strength (Natraja *et al.*, 2005). The absorption coefficient across a portion of the spectrum can be reordered into a monotonically increasing function called the k-distribution function (Modest & Zhang, 2002).

When the k-distribution method is used within a narrow band model, the longwave spectrum $k(\nu)$, is divided into N intervals ν_i , each having a smaller range of $k(\nu)$ values (Mlawer *et al.*, 1995). This procedure treats each subinterval in an equivalent manner as a spectral point is treated in a monochromatic radiative transfer method (Mlawer *et al.*, 1997). The division of these intervals is determined by the fact that they must be large enough to contain a significant amount of absorption lines associated with a particular absorber and also, they must be small enough so that the Planck function $B_\nu(T)$ can be treated as constant and equal to $\bar{B}_i$ across the band (Petty, 2004). The creation of the k-distribution involves assigning each absorption coefficient $k(\nu)$ a value g (0-1) that represents the fraction of the absorption coefficients in the band smaller than the representative $k(\nu)$. The k -values are re-arranged into ascending order transforming the spectrum into a smooth monotonically increasing function of the absorption coefficient, representing a cumulative k-distribution, g , defining a mapping, $\nu - g$ (Mlawer *et al.*, 1995). This can be seen in figure (3.1), which shows one such mapping, where the absorption coefficients for the spectral range (30-700 cm^{-1}) undergo a transformation into g -space figure (3.1b). The effect of this reordering is simply a rearrangement of the sequence of terms in the integral over wavenumber in the radiative transfer equations (Mlawer *et al.*, 1997).

By integrating over $k(g)$ instead of the complicated $k(\nu)$ we can replace the

3.3 k-distribution Method

3.4 Correlated k-distribution Method

3.4.1 Correlated k theory

The k-distribution method discussed above assumes a homogeneous path (one over which the temperature and pressure remain the same). This means the method would only suit very short paths through the atmosphere. This led to an extension of the k-distribution method to account for vertical paths in the

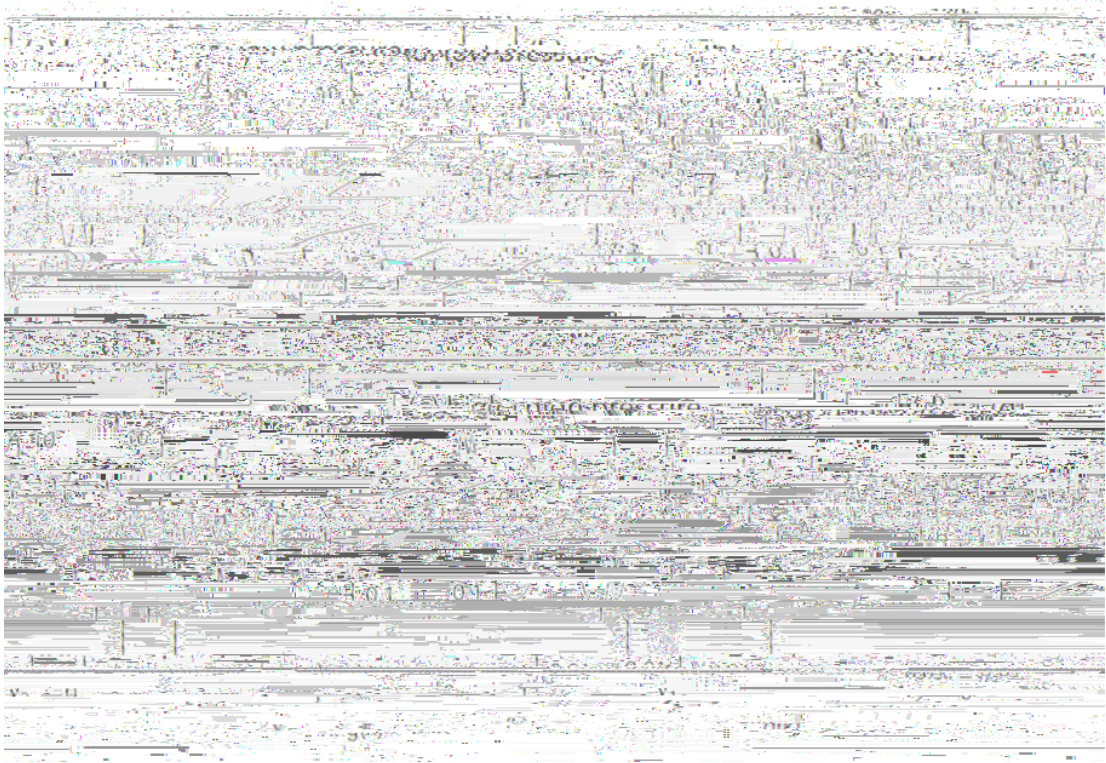


Figure 3.2: The k-distribution method and its extension to the correlated k method. (a) Absorption spectrum at low pressure, (b) sorting k to increase monotonically to form the function $g(k)$. (c) the same as for a and b, however at higher pressure, so are affected by pressure broadening (Petty, 2004)

3.4 Correlated k-distribution Method

homogeneous path (Petty, 2004).

$$T(u) = \exp \left(- \int_0^u k(g; u^0) du^0 \right) \quad (3.4)$$

This equation calculates the transmittance for a specific value of gu

3.5 Rapid Radiative Transfer Model (RRTM)

different frequencies. This can be seen in (3.2), the horizontal dotted line corresponds to a single value of k , yet has many values of frequency associated with it. In the CK Method, the absorption coefficient at all of these frequencies is associated to a single value of g (depicted by the intersection of the dotted line with the $k(g)$ curve figure (3.2b)).

Figure (3.2c) shows the absorption spectrum for a level higher in the atmosphere and its corresponding $k(g)$ mapping is represented in figure (3.2d). The absorption spectrum is affected by pressure broadening leading to a smoother spectrum with less extreme values of k . As a result, the spectral elements that contribute to a subinterval of the k -distribution for one homogeneous layer will not be mapped to the corresponding subinterval for a different atmospheric layer (Mlawer *et al.*, 1997). This leads to a different $k(g)$ distribution at the higher pressure. The newly determined frequencies at higher pressure, determined by the choice of k at lower pressure are not far off those frequencies at the lower pressure and this method does provide results within accuracy restraints.

The CKD method overestimates and underestimates the absorption in the vicinity of maximum and minima respectively. As a result, the CKD method may overestimate the spectral mean transmittance, leading to a more transparent atmosphere (Fu & Liou, 1992). The method does calculate fluxes and heating rates to errors of less than 1% (Petty, 2004) and is three orders of magnitude less computational power is required than LBL calculations (Petty, 2004).

3.5 Rapid Radiative Transfer Model (RRTM)

RRTM stands for the Rapid Radiative Transfer Model (Clough *et al.*, 2005). Modelled molecular absorbers are water vapour, carbon dioxide, ozone, nitrous oxide, methane, oxygen, nitrogen, and halocarbons. The model is accurate and fast, using the correlated- k method in its computation. RRTM divides the LW spectral region into 16 bands chosen for their homogeneity and radiative transfer properties (Mlawer *et al.*, 1995) (Clough *et al.*, 2005). There are a number of areas which can be explored to improve the speed of the model. The number

3.5 Rapid Radiative Transfer Model (RRTM)

of subintervals into which some of the bands are divided could be reduced and combining separate spectral regions which have similar absorbing properties into single spectral bands could help reduce the time taken for the model to run (Mlawer *et al.*, 1997).

(Fu & Liou, 1992) explored the optimum number of g values suitable within each band. They found that the number of quadrature points to provide sufficient accuracy could vary from 1 (for weak absorbing bands) to 10 (for strong absorption bands).

Each spectral band in the RRTM is broken into 16 intervals with 7 intervals lying between $g = 0.98$ and $g = 1.0$. This modified quadrature spacing is done to better determine the cooling rate where high values of $k(g)$ are associated with the centres of the spectral lines in the band (Mlawer *et al.*, 1995). The k -distribution is divided into subintervals of decreasing size with respect to g , with high resolution towards the upper end of the distribution. This arrangement allows accurate determination of middle atmosphere cooling rates while preserving the speed of the model (Mlawer *et al.*, 1997). At different atmospheric levels there is a limited range of absorption coefficient values that provide the main contribution to the cooling rate. Therefore when g approaches 1, only a small fraction of the k -distribution will have contributed due to the rapid increase in the function $k(g)$ at the high end of the distribution, which is why a high resolution is required near $g=1$ (Mlawer *et al.*, 1997). This high resolution in space near $g=1$ is difficult to achieve while maintaining speed of execution.

(Mlawer *et al.*, 1997) considers three main points when determining the number of spectral bands to use in RRTM:

Each spectral band can have at most two species with substantial absorption.

The range of values of the Planck function in each band cannot be extreme.

The number of bands should be minimal.

3.5 Rapid Radiative Transfer Model (RRTM)

An absorbing gas which dominates the absorption within a certain spectral band is termed a 'key species' and it is treated in more detail than other species in the bands that have smaller but still important absorption. These are referred to as 'minor species'. Table 3.1 presents the spectral bands of RRTM in the LW region (Mlawer *et al.*, 1997).

3.5 Rapid Radiative Transfer Model (RRTM)

Species Implemented in RRTM				
Band Number	Wavenumber Range, cm ⁻¹	Lower Atmosphere		Middle/Upper Atmosphere
		Key Species	Minor Species	Key Species
1	10-250	H ₂ O		H ₂ O
2	250-500	H ₂ O		H ₂ O
3	500-630	H ₂ O, CO ₂		H ₂ O, CO ₂
4	630-700	H ₂ O, CO ₂		13.65, O ₃
5	700-820	H ₂ O, CO ₂	CCl ₄	CO ₂ , O ₃
6	820-980	H ₂ O	CO ₂ , CFC-11, CFC-12	...
7	980-1080	H ₂ O, O ₃	CO ₂	O ₃
8	1080-1180	H ₂ O	CO ₂ , CFC-12, CFC-22	O ₃
9	1180-1390	H ₂ O, CH ₄		CH ₄
10	1390-1480	H ₂ O		H ₂ O
11	1480-1800	H ₂ O		H ₂ O
12	1800-2080	H ₂ O, CO ₂		...
13	2080-2250	H ₂ O, N ₂ O		...
14	2250-2380	CO ₂		13.65
15	2380-2600	CO ₂ , N ₂ O		...
16	2600-3000	H ₂ O, CH ₄		...

Table 3.1: RRTM Bands ([Mlawer et al., 1997](#))

3.5.1 Accuracy

The LW accuracy of RRTM is 0.6 Wm^{-2} (relative to the LBLRTM) for the net flux in each band at all altitudes with a total error of less than 1.0 Wm^{-2} at any altitude. It recorded an error of 0.07 K d^{-1} for total cooling rate error in the troposphere and lower stratosphere and 0.75 K d^{-1} in the upper stratosphere (Mlawer *et al.*, 1997).

Results of timing tests for RRTM indicate that computing the fluxes and cooling rates for a 51-layer atmosphere which includes the performance of 256 (16 bands x 16 subintervals) upward and downward radiative transfer calculations and, therefore, the computation of 256 optical depths per layer, takes 0.06 s on a SPARC server 1000. This compares favourably with the other rapid radiative transfer models (Mlawer *et al.*, 1995). A further indication of the computational efficiency of the model is that these radiative transfer operations in RRTM take 1.8 times the amount of time needed to perform $51 \times 16 \times 16$ exponentials. The speed and the accuracy of this model makes it suitable for use in GCMs (Mlawer *et al.*, 1997).

3.6 Full-Spectrum Correlated k-distribution Method (FSCK)

3.6.1 FSCK Method theory

The full-spectrum correlated-k distribution method (FSCK) is similar to the correlated k-distribution method (Li & Modest, 2002). In the FSCK method the absorption coefficients, k , are again sorted into smooth, monotonically increasing cumulative k-distributions, which enables the spectrum to be calculated with fewer calculations. The correlated k-distribution method is limited by the fact that the Planck function must be constant over the spectral range of the absorption coefficients. In the FSCK method, there is no restriction on the Planck function eliminating the need for a constant Planck function across each spectral

3.6 Full-Spectrum Correlated k-distribution Method (FSCK)

region (Pawlak *et al.*, 2004). As a result, in the FSCK approach, spectral bands can now be large, even as big as the full spectrum (Pawlak *et al.*, 2004).

Note that to eliminate the requirement that the Planck function must be constant over the spectral interval being sorted, the k-distribution is redefined using the Planck function as a weighting function. This means that the k-distribution gains temperature dependence through the Planck function (Pawlak *et al.*, 2004). (Stull, 1989) showed that an effective Planck function (the integral of the Planck function over wavenumbers which contribute to absorption in a particular range of g) can be used to replace the Planck function in the radiative transfer equations.

By eliminating the necessity for multiple spectral bands, the total number of calculations can be reduced substantially without losing significant accuracy relative to LBL calculations (Pawlak *et al.*, 2004).

Since FSCK requires quadrature over a single monotonically increasing function and needs about 10 quadrature points, while LBL calculations require about 1 million quadrature points, the FSCK method will greatly speed up the calculations. (Modest & Zhang, 2002) states that FSCK calculations required less than 0.05 seconds to perform defined calculations while the LBL calculations required 25 minutes.

Chapter 4

Method

4.1 Complex Radiative Transfer Chapter 4

4.1 Complex Radiative Transfer with Reynolds Decomposition

$$dF = d\bar{F} + dF^0 \quad (4.6a)$$

$$= \bar{\quad} + \quad^0 \quad (4.6b)$$

$$F = \bar{F} + F^0 \quad (4.6c)$$

$$B = \bar{B} + B^0 \quad (4.6d)$$

We begin by taking (4.5) and substituting into it the Reynolds decomposed versions of the variables found in (4.6):

$$d\bar{F} + dF^0 = D \bar{\quad} + \quad^0 \bar{F} + F^0 \bar{B} B^0 dz: \quad (4.7)$$

We can then multiply out the brackets:

$$d\bar{F} + dF^0 = D \bar{\quad} \bar{F} + \bar{\quad} F^0 \bar{B} \bar{B}^0 + \quad^0 \bar{F} + \quad^0 F^0 \quad^0 \bar{B} \quad^0 B^0 dz: \quad (4.8)$$

We now average over the whole spectrum, using rules 4, 5 and 6 from section 2.5, so any term with only one prime will cancel leaving

$$d\bar{F} = D \bar{\quad} \bar{F} \bar{B} + \overline{\quad^0 F^0} \overline{\quad^0 B^0} dz \quad (4.9)$$

This equation contains a covariance term which in principle we know $\overline{\quad^0 B^0}$, the covariance of the known spectral variation of the Planck Function with the known variation of the absorption coefficient.

the equation also contains a term we do not know $\overline{\quad^0 F^0}$. We can derive an expression for it by subtracting equation (4.9) from (4.8)

$$d\bar{F} + dF^0 - d\bar{F} = D \bar{\quad} \bar{F} + \bar{\quad} F^0 \bar{B} \bar{B}^0 + \quad^0 \bar{F} + \quad^0 F^0 \quad^0 \bar{B} \quad^0 B^0 dz - D \bar{\quad} \bar{F} \bar{B} + \overline{\quad^0 F^0} \overline{\quad^0 B^0} dz: \quad (4.10)$$

4.5

4.1 Complex Radiative Transfer with Reynolds Decomposition

$$dF^0 = D + \overline{F^0} \overline{B^0} + \overline{F^0} + \overline{F^0} \overline{B^0} - \overline{F^0} \overline{B^0} - \overline{F^0} + \overline{B^0} dz: \quad (4.11)$$

We want an equation for $d\overline{F^0}$ to predict $\overline{F^0}$. Using the chain rule we can expand $d\overline{F^0}$

$$d\overline{F^0} = \overline{dF^0} + \overline{F^0} d\overline{B^0}: \quad (4.12)$$

We are treating the extinction coefficient as constant along a path so $d\overline{B^0} = 0$

$$d\overline{F^0} = \overline{dF^0}: \quad (4.13)$$

We can now times (4.11) by $\overline{F^0}$ and take the average over the spectrum.

$$\overline{dF^0} = d\overline{F^0} = D - \overline{F^0} \overline{B^0} + \overline{F^0} \overline{B^0} - \overline{F^0} \overline{B^0} - \overline{F^0} \overline{B^0} dz: \quad (4.14)$$

The last two terms in (4.11) disappear when the average is taken and we obtain (4.14) which again contains an unknown, this time of a higher order $\overline{F^0}$

We can follow the same procedure again to obtain an equation for the new unknown.

$$d\overline{F^0} = \overline{dF^0} = D - \overline{F^0} \overline{B^0} + \overline{F^0} \overline{B^0} - \overline{F^0} \overline{B^0} - \overline{F^0} \overline{B^0} dz: \quad (4.15)$$

Now we again have another unknown this time of higher order $\overline{F^0}$ again this is a closure problem. We could keep on deriving equations for the unknowns, however, each time we derive a new equation, it will have a new unknown of one higher order than that of the term we are trying to derive. This is known as the closure problem [Stull \(1989\)](#). We overcome this problem, by parameterizing the unknown term at a certain stage where we require a certain accuracy.

4.1 Complex Radiative Transfer with Reynolds Decomposition

So up to third order accuracy we can define F by these three equations

$$\frac{d\bar{F}}{dz} = D(\bar{F} - \bar{B} + \overline{\theta F^0} + \overline{\theta B^0}) \quad (4.16a)$$

$$\frac{d\overline{\theta F^0}}{dz} = D(\overline{\theta^2 F} - \overline{\theta^2 B} + \overline{\theta F^0} + \overline{\theta^2 F^0} - \overline{\theta B^0} - \overline{\theta^2 B^0}) \quad (4.16b)$$

$$\frac{d\overline{\theta^2 F^0}}{dz} = D(\overline{\theta^3 F} - \overline{\theta^3 B} + \dots)$$

4.2 Simple Radiative Transfer with Reynolds Decomposition

We will simplify the above example even further. In the case where the Planck Function B equals zero (no emission) and the Diffusivity Factor D equals one (Radiation travels in only one direction). Equation (

4.2 Simple Radiative Transfer with Reynolds Decomposition

new approximation, this time of order 3.

$$d \overline{\tau^2 F_0} = \overline{\tau^3} \overline{F} + \overline{\tau^2 F_0} + \overline{\tau^3 F_0} dz \quad (4.22)$$

Now $\overline{\tau^3 F_0}$ is an unknown.

Up to third order accuracy we can define F by these three equations:

$$\frac{d\overline{F}}{dz} = -\overline{F} - \overline{\tau F_0} \quad (4.23a)$$

$$\frac{d \overline{\tau F_0}}{dz} = \overline{\tau^2} \overline{F} - \overline{\tau F_0} - \overline{\tau^2 F_0} \quad (4.23b)$$

$$\frac{d \overline{\tau^2 F_0}}{dz} = \overline{\tau^3} \overline{F} - \overline{\tau^2 F_0} - \overline{\tau^3 F_0} \quad (4.23c)$$

If we set $F = \overline{F}$, $G = \overline{\tau F_0}$, $H = \overline{\tau^2 F_0}$ and $I = \overline{\tau^3 F_0}$ we obtain:

$$\frac{dF}{dz} = -F - G \quad (4.24a)$$

$$\frac{dG}{dz} = \overline{\tau^2} F - G - H \quad (4.24b)$$

$$\frac{dH}{dz} = \overline{\tau^3} F - H - I \quad (4.24c)$$

It is clear that [4.24](#)

4.3 Model Description

Again we have a closure problem as described in the previous section. The

4.3 Model Description

of x , so a normal distribution of x willtionwilltg8wil8b

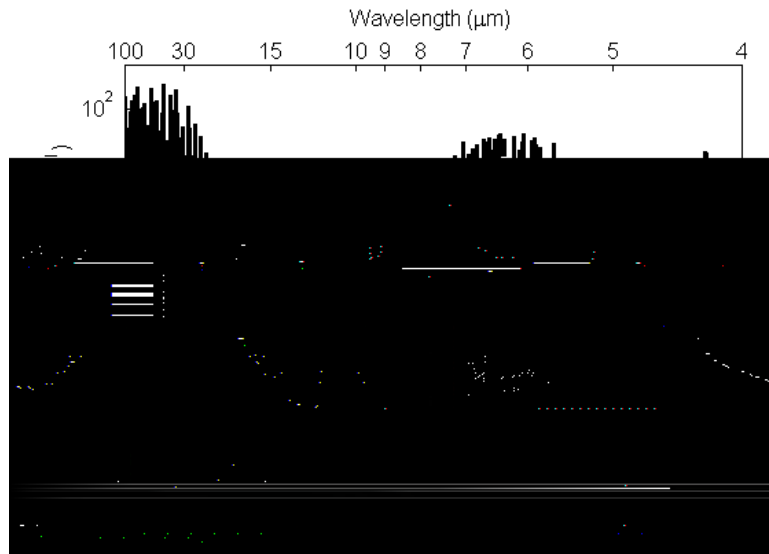


Figure 4.2: Extinction coefficient spectrum due to CO_2 , H_2O and O_3 in the IR region

The extinction coefficient is measured in m^{-1} as it is a measure of the molecular absorption cross-section, per m^3 of air.

The external file also includes data for the Planck function over the wavelength spectrum shown in figure (4.3).

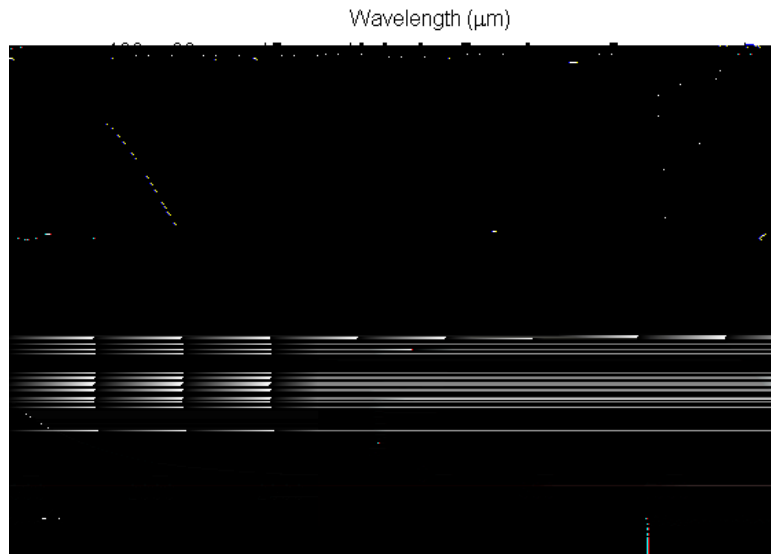


Figure 4.3: Variation of Planck Function with wavelength and wavenumber in over IR region

Chapter 5

Results

5.1 Basic Model

5.1.1 Normal Distribution

Figure (5.1) shows the transmittance along a homogeneous path through the atmosphere. The thick black line represents the true transmittance and is equivalent to the LBL calculations. These graphs were obtained using 5000 points in z and a space step of, $dz = 20$.

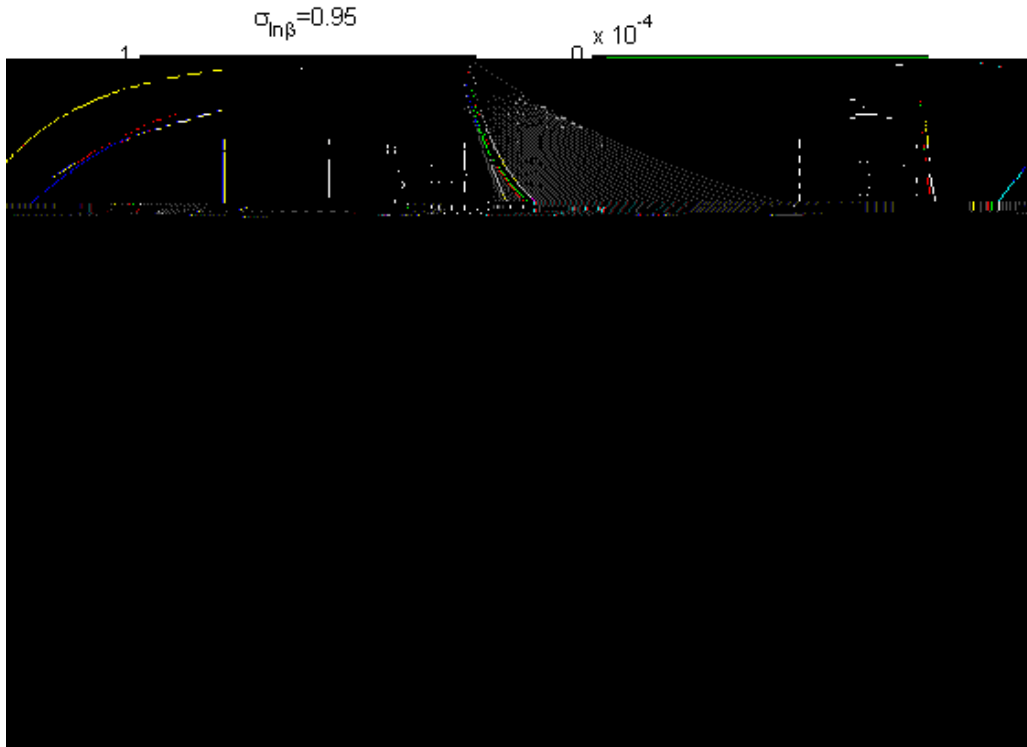


Figure 5.1: Blue = numerical first order transmittance, red = second order numerical transmittance, green = numerical third order transmittance. Normal distribution, $\sigma_{\ln \beta} = 0.95$.

The blue line represents the numerical first order radiative transfer when the fluctuating terms are averaged across the whole spectrum. The scaling factor is 0.357 which determines the spread of the absorption coefficients, this results in a standard deviation, σ , of 0.95. The red line is the second order numerical transmittance and the green line is the numerical third order transmittance. The many grey lines represent the raw transmissions of each absorption coefficient. There are 61 absorption coefficients in the spectrum. It is therefore clear from figure 5.1 that the absorption coefficient spectrum is normally distributed represented by figure 5.2.

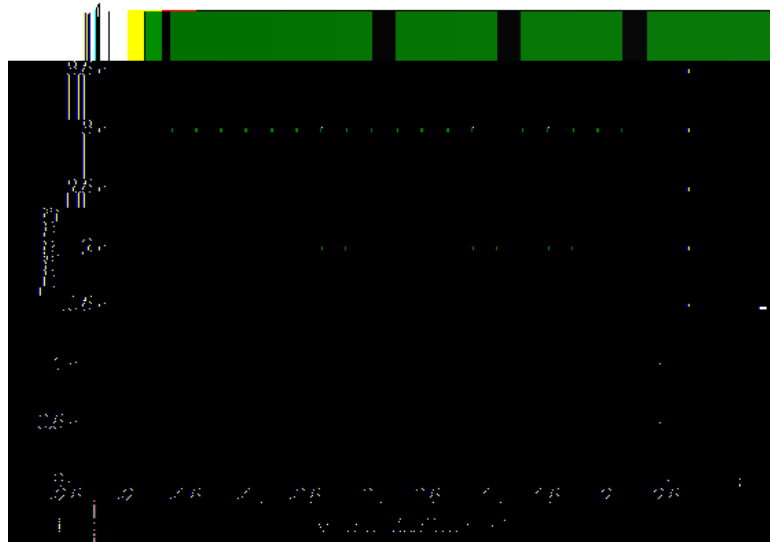


Figure 5.2: Histogram of extinction coefficients, normally distributed.

Figure 5.3

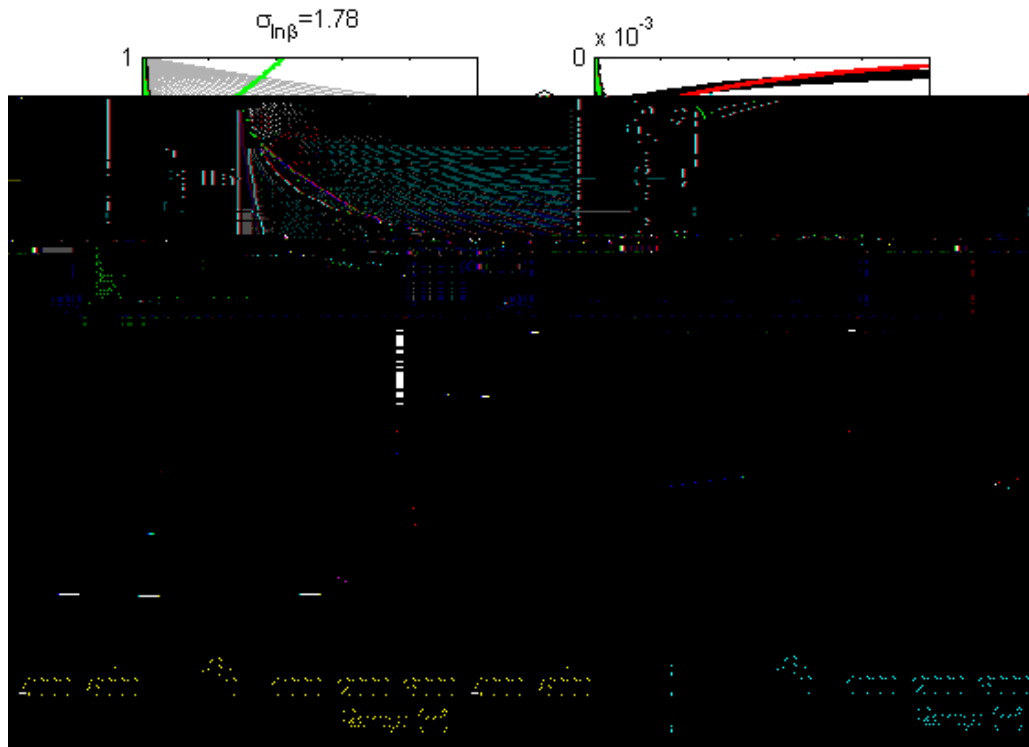


Figure 5.3: Blue = numerical first order transmittance, red = second order nu-

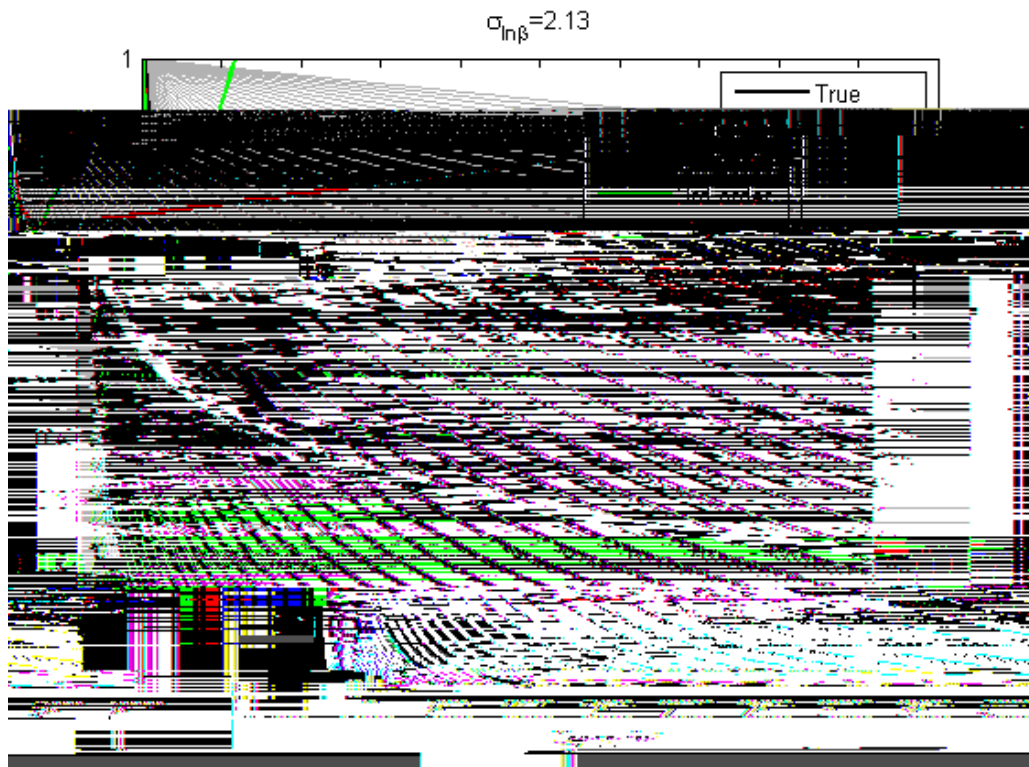


Figure 5.4: Blue = numerical first order transmittance, red = second order numerical transmittance, green = numerical third order transmittance. Normal distribution, $\sigma_{\ln\beta}=2.13$.

The variation in covariance has not been plotted as the third order (green)

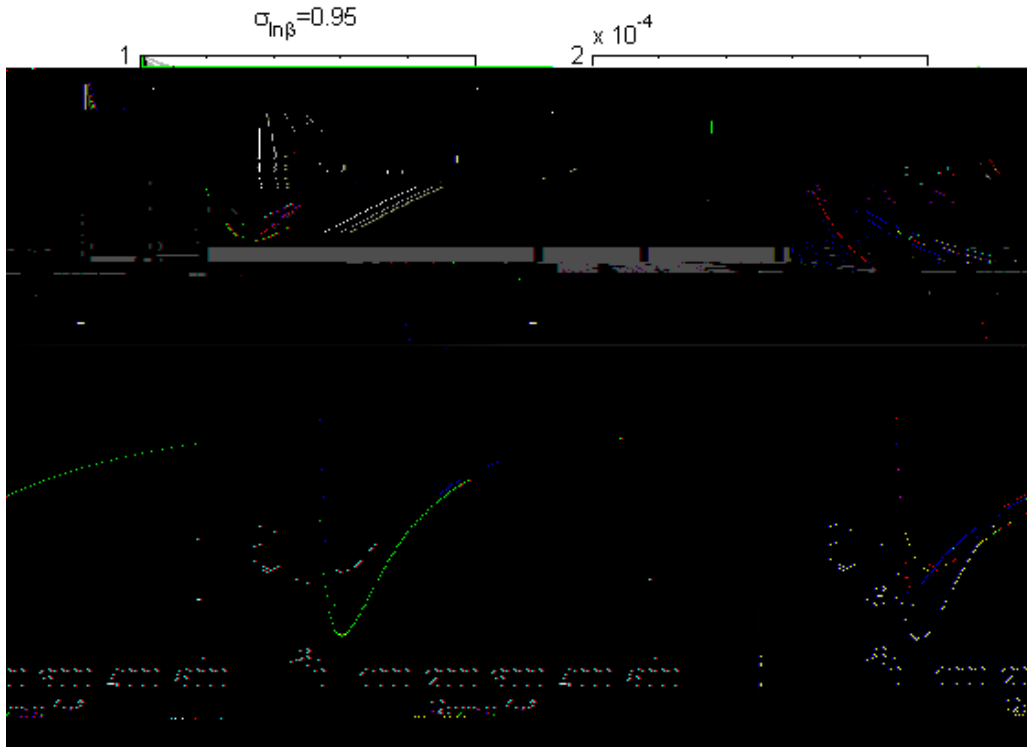


Figure 5.5: Blue = numerical first order transmittance, red = second order numerical transmittance, green = numerical third order transmittance. Gaussian distribution, $\sigma_{\ln\beta}=0.95$.

Figure 5.5 has a scaling factor of one to achieve the standard deviation $\sigma_{\ln\beta} = 0.95$ and also has 61 absorption coefficients in the spectrum. It is clear from the raw transmission for each individual absorption coefficient (faint grey lines) that the distribution is Gaussian, represented in figure 5.6.

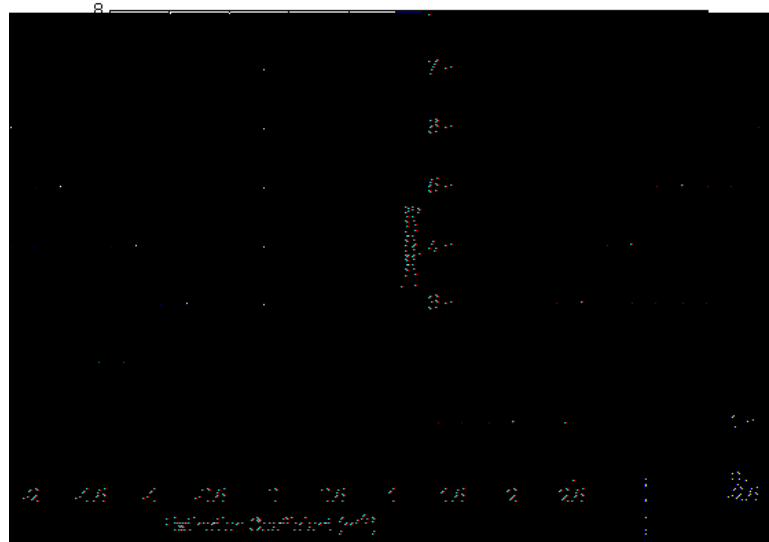


Figure 5.6: Histogram of extinction coefficients, Gaussian distributed.

It appears that for the same standard deviation in a normal distribution, figure 5.1 and a Gaussian distribution, figure 5.5, the Gaussian distribution is more accurate at approximating the transmittance along the path. As the distribution is Gaussian however, each time the model is run, the distribution changes. Figure 5.7 has the same standard deviation, $\sigma = 0.95$, as figure 5.5, yet has produced different approximations for each of the first, second and third order transmissions.

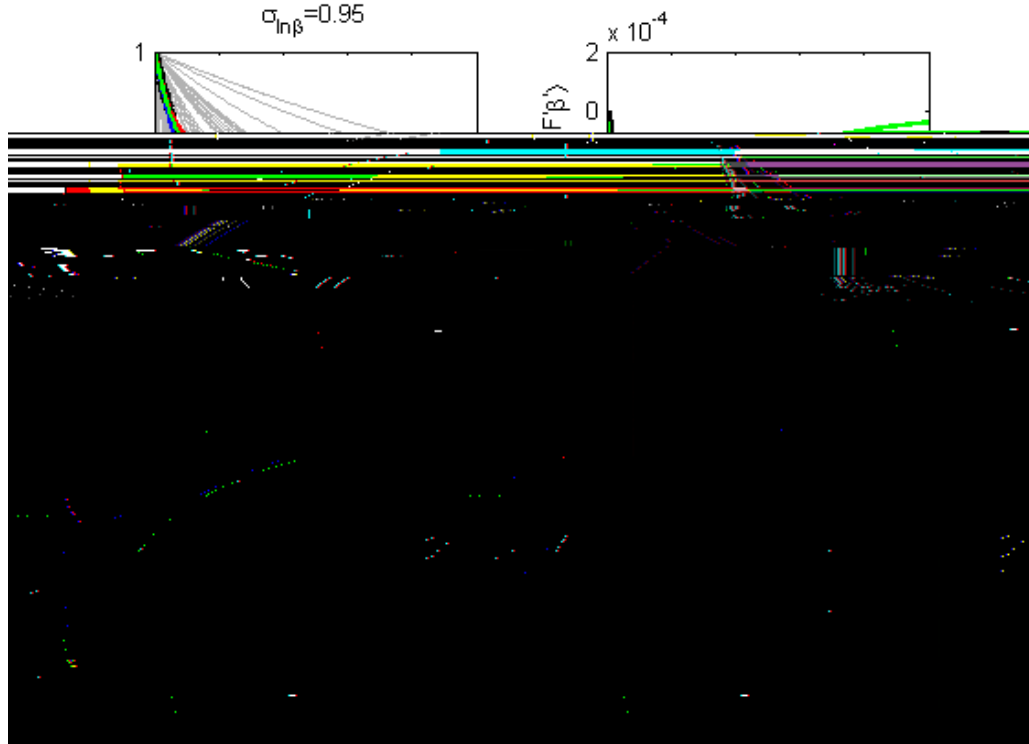


Figure 5.7: Blue = numerical first order transmittance, red = second order numerical transmittance, green = numerical third order transmittance. Gaussian distribution, $\sigma_{\ln \beta} = 0.95$.

Figure 5.8 has a scaling factor of 1.89 to achieve the standard deviation the same as Figure 5.3 yet the second order approximation has blown up very quickly and the third order transmission drops quickly to zero.

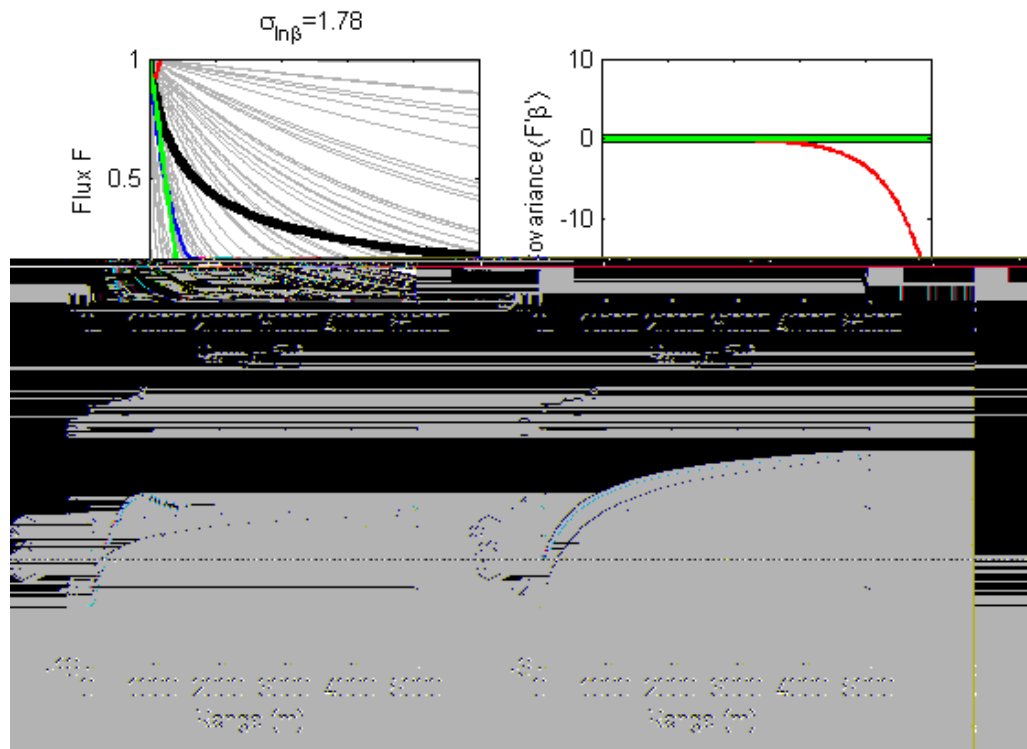


Figure 5.8: Blue = numerical first order transmittance, red = second order numerical transmittance, green = numerical third order transmittance. Gaussian distribution, $\sigma_{\ln \beta} = 1.78$. **Simple two band model**

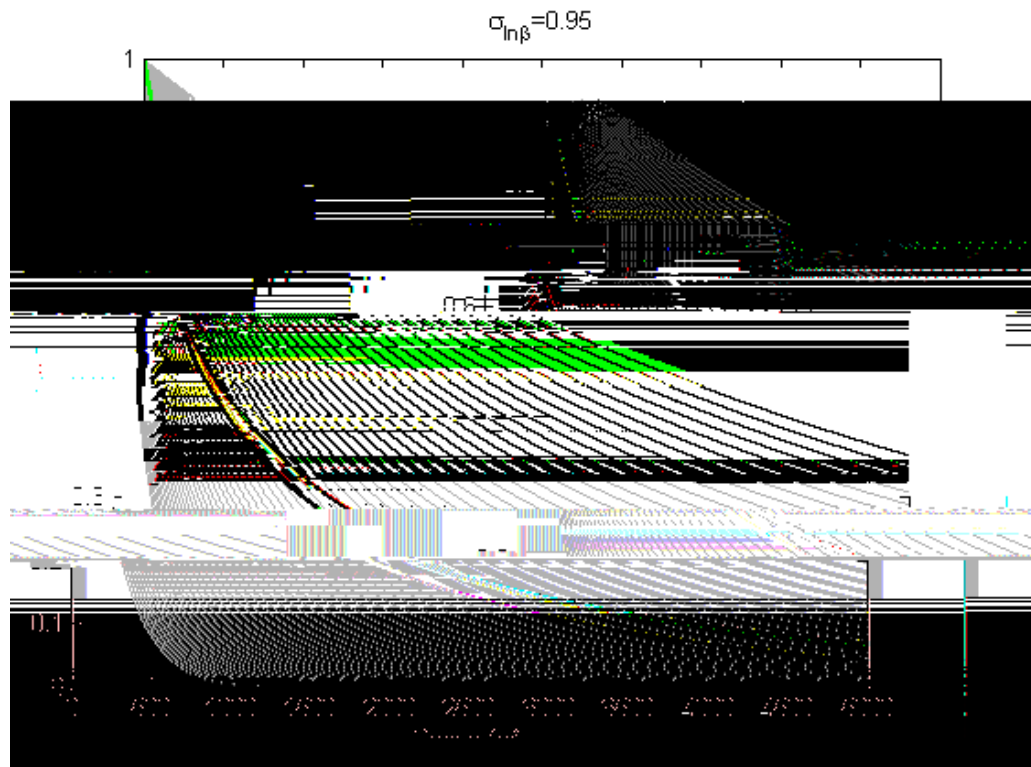


Figure 5.9: Two band model. Blue = numerical first order transmittance, red = second order numerical transmittance, green = numerical third order transmittance. Normal distribution, $\sigma_{\ln \beta} = 0.95$.

5.2.2 Gaussian

Figure 5.10 has a Gaussian distribution of absorption coefficients and has the same standard deviation as the Gaussian distributed figure 5.5.

5.3 Equal weight band model

5.3.1 Atmospheric data

In the absorption coefficient spectrum, there are 9610 values plotted in figure 4.2. Figure 5.11 was produced sampling every single absorption coefficient. The model takes a while to run, sampling 9610 points. Figure 5.11 shows the root mean squared (RMS) accuracy of the transmission of the first second and third order numerical approximations compared to the true transmission as the number of bands in the model is increased.

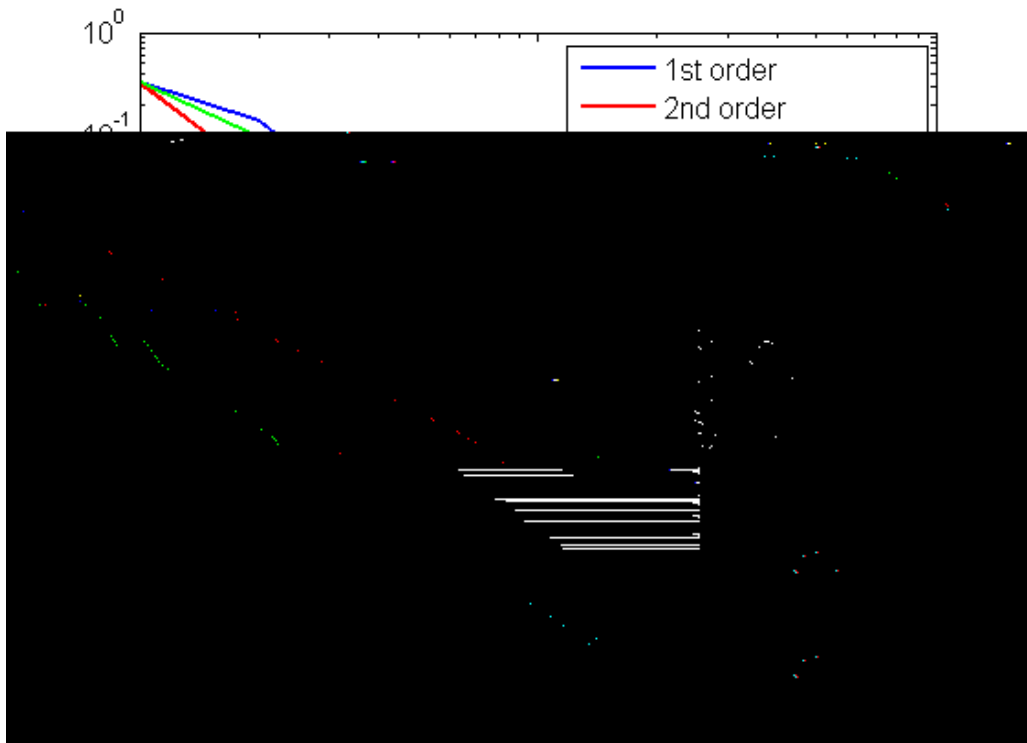


Figure 5.11: Equally weighted bands. Blue = numerical first order transmittance RMS error, red = second order numerical transmittance RMS error, green = numerical third order transmittance RMS error, for atmospheric data.

We can however sample a smaller number of extinction coefficients such as every 30 shown in figure 5.12.

5.4.1 Atmospheric data

Figure 5.13 was produced sampling every single absorption coefficient. Figure 5.13 shows the root mean squared (RMS) accuracy of the transmission of the first, second and third order numerical approximations compared to the true transmission as the number of bands in the model is increased.

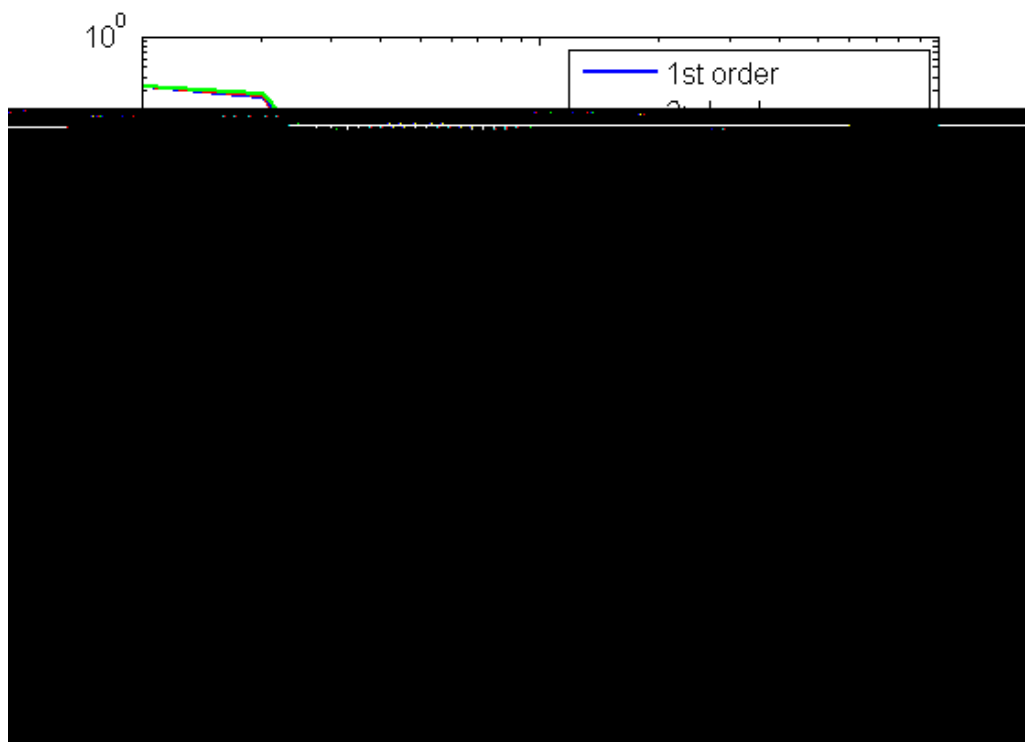


Figure 5.13: Weighted bands. Blue = numerical first order transmittance RMS error, red = second order numerical transmittance RMS error, green = numerical third order transmittance RMS error, for atmospheric data.

Figure 5.14 was produced sampling every 20th absorption coefficient. The model samples 480 points. Already the first order approximation is beginning to blow up. So with this technique we have to carry out a finer sample of absorption coefficient.

5.4 Weighted average band model

5.4 Weighted average band model

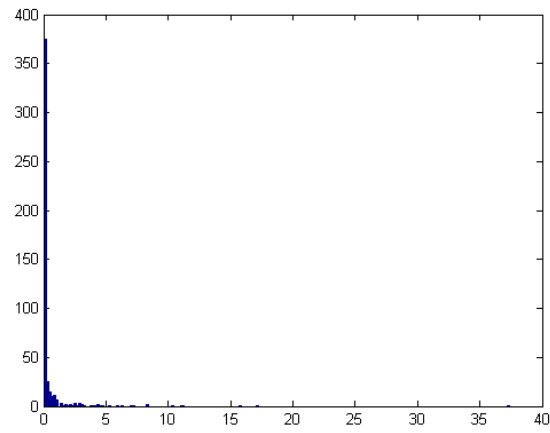


Figure 5.15: Histogram of atmospheric data extinction coefficients.

Chapter 6

Discussion

For a normal distribution of absorption coefficients and a standard deviation of, $\sigma = 0.95$, figure [5.1](#)

are generated randomly, a different spectrum is produced each time. This can lead to a very different distribution with the same standard deviation, dramatically affecting the ability to produce accurate results. This is shown by comparing figure 5.7 with figure 5.5, which have the same standard deviation, 0.95, yet they have both produced different approximations for each of the first, second and third order transmissions.

In section 5.2, it is clear that by simply splitting the spectrum into two bands, the accuracy of all the three different schemes are better at approximating the true transmission. Both the normal and Gaussian distributions have improved their accuracy compared to performing the calculation across the whole spectrum.

Figure 5.9 samples all 9610 absorption coefficients in the atmospheric data file and represents the accuracy of the numerical schemes compared to the true transmission when the number of bands (each with the same number of absorption coefficients) is increased from one to 100. An accuracy of 10^{-6} Wm^{-2} is achieved for 100 bands. This is an acceptable error for the scheme.

It is not essential to sample the absorption coefficient at every value. Relatively good accuracy is achievable by sampling every 10 or 20, figure 5.12. In sampling the absorption spectrum more sparsely we can reduce the time taken to compute the transmission. The numerical third order approximation is the quickest to increase in accuracy as the number of bands are increased, however at around 15 bands, the accuracy levels off, the reasons why need to be investigated. The numerical second order approximation is the most accurate reaching a RMS error of 10^{-6} Wm^{-2} by 100 bands. The first order approximation reaches an RMS error of 10^{-4} Wm^{-2} by 100 bands.

We would expect the band model in section 5.4, which has an equal fraction of the dynamic range to perform better in terms of accuracy, as a relatively larger number of lesser important (smaller absorption coefficients) are treated all together than when each band has an equal number of absorption coefficients. This means that the more important, larger absorption coefficients are treated across more bands so are considered in greater detail than in the equally weighted band model. This would appear to be disputed however as figures 5.11 and 5.12

are producing results to a greater accuracy than the equally weighted band model, figures [5.13](#) and [5.14](#).

It is clear that the models discussed have difficulty in handling a large range in extinction coefficient, this is why bands are so important and have been used throughout the development of radiative transfer models. They enable a smaller range of absorption coefficients to be considered within each band and therefore

Chapter 7

Conclusions

The next step in the development and testing of a Reynolds averaged radiative transfer model would be to include radiation travelling in different directions, $D=1.66$ and to also include emission along the path, therefore having to include the planck function. This would involve solving the systems of equations in (4.17). Then we would like to account for scattering. We would also like to account for an inhomogeneous path, where pressure and temperature change, i.e. vertically through the atmosphere, so the absorption spectrums of the gases will change due to pressure broadening.

The next step might be to develop the maths for application to shortwave atmospheric radiative transfer and ultimately to develop a radiative transfer module that may be used in operational numerical weather prediction or GCMs. Comparisons between the newly developed method and current models would be required to determine the relative accuracy of the model. The use of Reynolds decomposition within the radiative transfer equations has the potential for improving the computational requirements of the radiative transfer code relative to traditional correlated k-distribution methods.

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