Mie Theory Calculations

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September 11, 2009

Contents

1   Introduction

In the article *Electromagnetic Scattering From a Sphere* [1] I introduced the basics of the Mie theory, as well as some extensive revision on electromagnetism and mathematics. At the end I pointed the reader to Warren Wiscombe's excellent code that performs the task, and recommended the use of this code. However, that of course didn't stop me writing my own, and this article describes the nitty gritty of actually performing any calculations.

However, there is difference between what is described here and what was done by Wiscombe in that we shall require the expansion of the scattering functions in Legendre polynomials. This can be done analytically instead of by using numerical integration.

For practical applications, it is common that an integration over some statistical distribution of particle distributions. It is often the case that this will be a skew distribution with a long tail containing large particle sizes. It is here the ripple-spike problem occurs as described in appendix-B of Wiscombe's NCAR report [2]. Because of this, we shall investigate the sources of any possible numerical instability.

At the end, a link is provided so that the reader can obtain my own code which calculates the Legendre coefficients given a modified gamma distribution of particle sizes.

2   The Mie Coefficients

We begin by recalling that the Mie a and b coefficients are at the heart of the calculation, and that for a sphere of radius a these were given by Here, $n \geq 1$, $x = ka$

$$a_n = \frac{\psi'_n(y)\psi_n(x) - m\psi_n(y)\psi'_n(x)}{\psi'_n(y)\xi_n(x) - m\psi_n(y)\xi'_n(x)}. \tag{1}$$

$$b_n = \frac{m\psi'_n(y)\psi_n(x) - \psi_n(y)\psi'_n(x)}{m\psi'_n(y)\xi_n(x) - \psi_n(y)\xi'_n(x)}. \tag{2}$$

Here, $n \geq 1$, $x = ka = 2\pi a/\lambda$, and $y = ma$ for a complex refractive index m. The functions ψ and χ arise in eqns.57-60 in [1]. Here, the ξ are similar to the Hankel functions in that $\xi = \psi - i\chi$. Generally, the series where the a_n and b_n are the coefficients shall be truncated at some value N.

We shall follow Diermendjian [3] and write

$$a_n = \frac{A_n(y)\psi_n(x) - m\psi'_n(x)}{A_n(y)\xi_n(x) - m\xi'_n(x)}. \tag{3}$$

$$b_n = \frac{mA_n(y)\psi_n(x) - \psi'_n(x)}{mA_n(y)\xi_n(x) - \xi'_n(x)}. \tag{4}$$

Here, we have kept the notation from [1] , and simply introduced

$$A_n(y) = \frac{\psi_n'(y)}{\psi_n(y)}.$$ (5)

or

$$A_n(y) = -\frac{n}{y} + \frac{J_{n-1/2}(y)}{J_{n+1/2}(y)}.$$ (6)

Furthermore, looking at the properties of the Bessel functions [4], it turns out to be convenient to introduce the functions

$$w_n(x) = \sqrt{\frac{\pi x}{2}}[J_{n+1/2}(x) - (-1)^n J_{-n-1/2}(x)].$$ (7)

When this done, the Mie coefficients are written as

$$a_n = \frac{\left(\frac{A_n(y)}{m} + \frac{n}{x}\right) Re\{w_n(x)\} - Re\{w_{n-1}(x)\}}{\left(\frac{A_n(y)}{m} + \frac{n}{x}\right) w_n(x) - w_{n-1}(x)}$$ (8)

$$b_n = \frac{\left(mA_n(y) + \frac{n}{x}\right) Re\{w_n(x)\} - Re\{w_{n-1}(x)\}}{\left(mA_n(y) + \frac{n}{x}\right) w_n(x) - w_{n-1}(x)}$$ (9)

3   The A_n

The first thing we shall consider is the behaviour of the A_n, the $J_{n+1/2}$ in the denominator of eqn.6 has many zeros as seen in Fig.1. So for a real refractive index for instance, and this will mean trouble. This is easily sidestepped in practice though.

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We suppose the A_n shall be stored in some array $BIGA$, along with this array we can have a boolean array *reciprocal* stored alongside it. So we use the approach

```
if(J_(n+1/2)* 100.0  > J_(n-1/2))then
            store  -n/y+J_(n-1/2)/J_(n+1/2) in BIGA
            store  false in reciprocal
        else
            store y J_(n+1/2)/(y J_(n-1/2)-nJ_(n+1/2) ) in BIGA
            store true in reciprocal

        if( reciprocal is false)then
            use eqns 8 and 9 to calculate a_n and b_n
        else
            use eqns 10 and 11 to calculate a_n and b_n
```

So if the denominator in eqn.6 is close to blowing up, we use

$$a_n = \frac{\left(\frac{1}{m} + \frac{n A_n^{-1}(y)}{x}\right) Re\{w_n(x)\} - A_n^{-1}(y) Re\{w_{n-1}(x)\}}{\left(\frac{1}{m} + \frac{n A_n^{-1}(y)}{x}\right) w_n(x) - A_n^{-1}(y) w_{n-1}(x)}, \tag{10}$$

$$b_n = \frac{\left(m + \frac{n A_n^{-1}(y)}{x}\right) Re\{w_n(x)\} - A_n^{-1}(y) Re\{w_{n-1}(x)\}}{\left(m + \frac{n A_n^{-1}(y)}{x}\right) w_n(x) - A_n^{-1}(y) w_{n-1}(x)}. \tag{11}$$

Of course,

$$A_n^{-1}(y) = \frac{y J_{n+1/2}(y)}{y J_{n-1/2}(y) - n J_{n+1/2}(y)} \tag{12}$$

follows immediately from eqn.6. So both the A_n and the reciprocals are stored in the same

array, with the boolean array telling us which formula for the Mie coefficients to use.
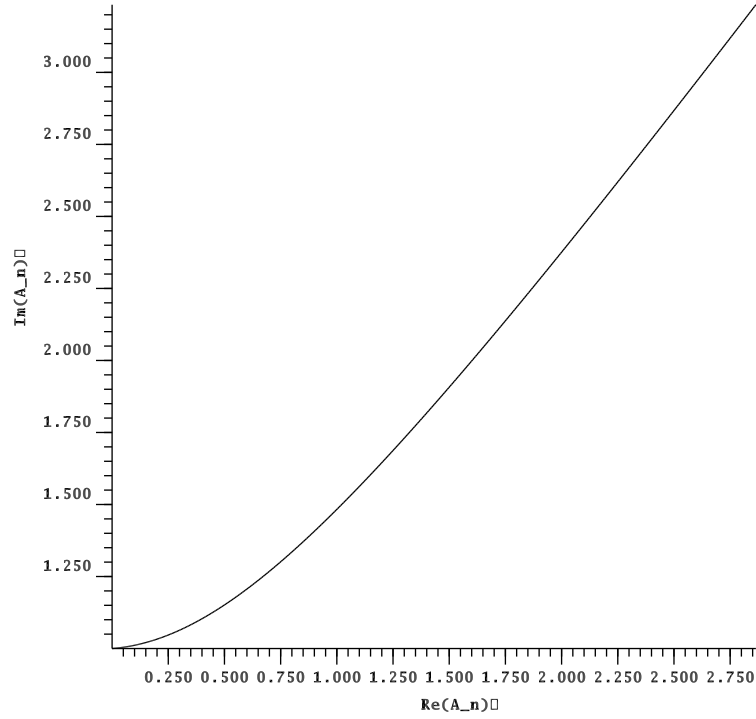

Figure 2: The real and imaginary parts of A_n are roughly $0 + i$ small n, and increase to roughly $2.75 + 3i$ as $n \to 500$.

Deirmendjian calculated the A_n by determining recurrence relations, and found that in the problem for $x = 62$, and $m = 1.28 - 1.37i$ that in using double precision arithmetic, the imaginary part of the A_n was unstable after $n > 60$ or so. However, modern library routines such as the amos library (which is also part of the slatec library) are highly accurate, robust, and efficient. In Fig.2 we plot the real and imaginary parts of $A_n(y)$ in this case for $n = 1 - 500$ for exactly the same problem. Here we have used eqn.6, and used the slatec zbesj function. We see there is no problem up to $n = 500$ at all. However, for this case, as n increases still further, both the real and imaginary parts of $J_{n+1/2}(y)$ become zero as far as double precision arithmetic is concerned.

So what should be done if, for large n, both $J_{n+1/2}$ and $J_{n+1+1/2}$ are returned as zero by zbesj? According to Abramowitz and Stegun §9.3 [4], for large orders

$$J_\nu(z) \sim \frac{1}{\sqrt{2\pi\nu}} \left(\frac{ez}{2\nu} \right)^\nu \tag{13}$$

Here, z is kept constant as ν increases Of course, for real arguments for instance, this does not appear to be of much use because of the oscillatory nature of the Bessel functions. For this reason, Deirmendjian concluded that "there was no practical way of making use of them

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in an exact machine program". However, in Fig.3 we plot the ratio $J_{n-1/2}(y)/J_{n+1/2}(y)$ with a real refractive index of $m = 1.28$.

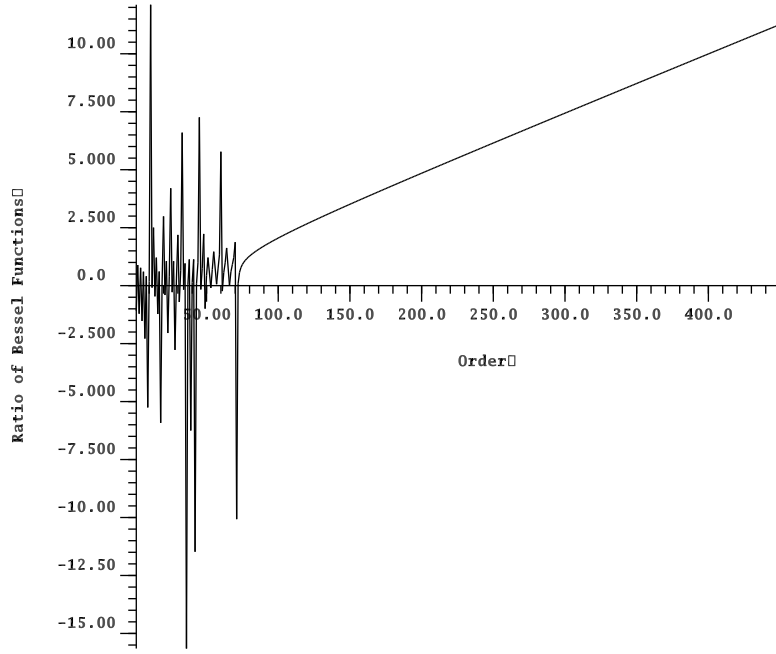Figure 3: $J_{n-1/2}(y)/J_{n+1/2}(y)$ as a function of n for m=1.28

When n becomes large enough, the oscillations die out, so can eqn.13 be of any use after all? We observe from eqn.13 that

$$\frac{J_\nu(y)}{J_{\nu+1}(y)} \sim \sqrt{\frac{\nu+1}{\nu}} \frac{2(\nu+1)}{ez} \left(\frac{\nu+1}{\nu}\right)^\nu \tag{14}$$

and in the above example, the approximation is out by about one percent for $n = 450$. If the imaginary part of the refractive index is large as in Deirmendjian's example above, there is a loss of of accuracy in the imaginary part of $A_n(y)$ of 3% or so in the cases looked at by the author.

However, if we cannot jump to greater precision arithmetic this small error must at least be more acceptable than truncating the Mie series too early. When n s large enough (depending on y) the errors in A_n will be dominated by the n/y term at any rate. This shall be discussed further below, where it is seen that as far as calculating the Mie a and b coefficients is concerned, it doesn't matter at all.

Before proceeding we note that as n is increasing and the absolute values of the Bessel functions in eqn.6 are heading to zero, the denominator may hit zero first, but it may not be a true zero of the actual Bessel functions. We must check the results of zbesj and find

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out at what point are all the Bessel functions are zero and use the asymptotic results after. Of course, it may be that once the Bessel functions are very small indeed, that the relative accuracy may be low anyway, so perhaps a switching at an earlier point will be better. For the cases studied by the author though, this does not seem to be the case.

4   The Denominator in a_n and b_n

One feasible problem in calculating the a_n and b_n is that the denominator could become zero, or be very close to zero. Physically, we know that the a_n and b_n are always finite, so that if this were the case, the numerator must also be zero or extremely small, in which case we must use L'Hospital's rule. Here we ask whether this is ever the case.

Now the w_n of equations 8 and 9 are given by

$$w_n(x) = \sqrt{\frac{\pi x}{2}}(j_n - iy_n) \tag{15}$$

where the j and y are the spherical Bessel functions. This might well give problems. There is a singularity at $x = 0$ so for large orders and small x the slatec routine dbesy just gives an overflow message and stops, however, the routine can be easily modified to give an error code return and we for a given x we can call the routine with a smaller maximum order and truncate there. Can we simply do this, or is the "small differences between large numbers" type error ever a problem?

Let us rewrite eqn.8 in a different form.

$$a_n = \frac{\left(\frac{A_n(y)}{m} + \frac{n}{x}\right)Re\{w_n(x)\} - Re\{w_{n-1}(x)\}}{y_n(x)\left[\left(\frac{A_n(y)}{m} + \frac{n}{x}\right)\left(\frac{j_n(x)}{y_n(x)} - i\right) - \left(\frac{j_{n-1}(x)}{y_n(x)} - i\frac{y_{n-1}(x)}{y_n(x)}\right)\right]}. \tag{16}$$

Now for small x we have

$$y_n(x) \approx \frac{1}{x^{n+1}} \times 1 \times 3 \times 5 \dots \times (2n - 1) \tag{17}$$

so

$$y_0(x) \approx \frac{1}{x},$$

$$y_1(x) \approx \frac{1}{x^2},$$

$$y_2(x) \approx \frac{1}{x^3} \times 3,$$

$$y_3(x) \approx \frac{1}{x^4} \times 15,$$

$$y_4(x) \approx \frac{1}{x^5} \times 105, \tag{18}$$

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and so on. So, we have $y_{n-1}(x)/y_n(x) \approx x/(2n-1)$ for $n = 1, 2, \dots$. The question is this, although $y_n(x)$ will be very large for small x and particularly for large order, can the absolute value of the quantity

$$f = \left[\left(\frac{A_n(y)}{m} + \frac{n}{x} \right) \left(\frac{j_n(x)}{y_n(x)} - i \right) - \left(\frac{j_{n-1}(x)}{y_n(x)} - i \frac{y_{n-1}(x)}{y_n(x)} \right) \right] \tag{19}$$

become very small? For small enough x we put

$$f \approx -i \left[\left(\frac{A_n(y)}{m} + \frac{n}{x} \right) - \frac{x}{2n-1} \right]. \tag{20}$$

So we might immediately observe that the reciprocal term in x dominates, particularly for large n, and we so the function f will certainly not vanish. However this assumes that $x << 1$, but the y_n can blow up for even $x >> 1$ because of the $(2n-1)!!$ term and we cannot assume that $x << 1$ at all. Now, if the imaginary part part of of A_n is non zero, then because all the other terms in eqn.20 are real, we cannot arrive at the situation where the absolute value of eqn.20 becomes zero. However, A_n is pure real if the refractive index is pure real.

The author has investigated the behaviour of the f, in regions where the approximation eqn.20 is valid, for a range of refractive indexes, x and to high order, and it seems that the function f doesn't vanish under any circumstances in this case. For large enough x the y_n are also oscillatory, but again, for all x and all orders the denominator doesn't vanish, so L'Hospital's rule is never required. So it would appear that truncating the Mie series at the first value of N for which dbesy does not return an error code will be fine for a given value x.

Not only this, but the if we truncate in this way, we find in practice that we never reach values of N such that the asymptotic expression for the A_n are needed.

5   Other Possible Sources of Error

Let us take a look at a_1 and b_1 as functions of x as in Fig.4 and Fig.5. First, as Deirmendjian states, that if the refractive index is pure real, that the a_n and b_n all lie on a circle, of radius $1/2$, and centred at $z = (1/2, 0)$ no matter what the value of x is. If the imaginary part the refractive index is non zero, then all values of a_n and b_n lie within this circle. Not only that, but as $x \to \infty$ the a_n and b_n tend to a circle with a given radius of convergence also

centred at $z = (1/2, 0)$.

Figure 4: The real and imaginary parts of a_1 for a refractive index 1.29-0.0427i

Figure 5: The real and imaginary parts of b_1 for a refractive index 1.29-0.0427i

The behaviour is complicated, but the circle of constant radius emerges clearly for large x, the same behaviour is seen in Figs.6 and 7 for a different refractive index, and in Fig.8
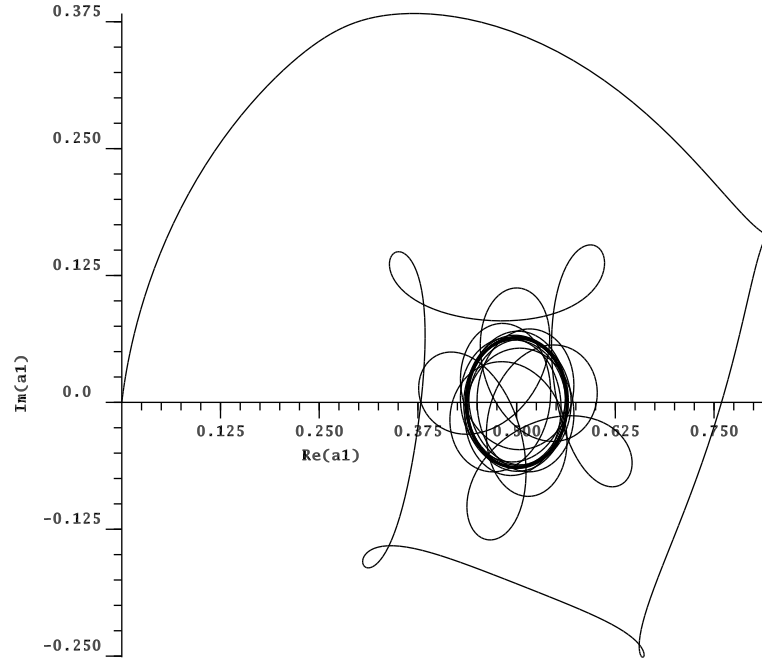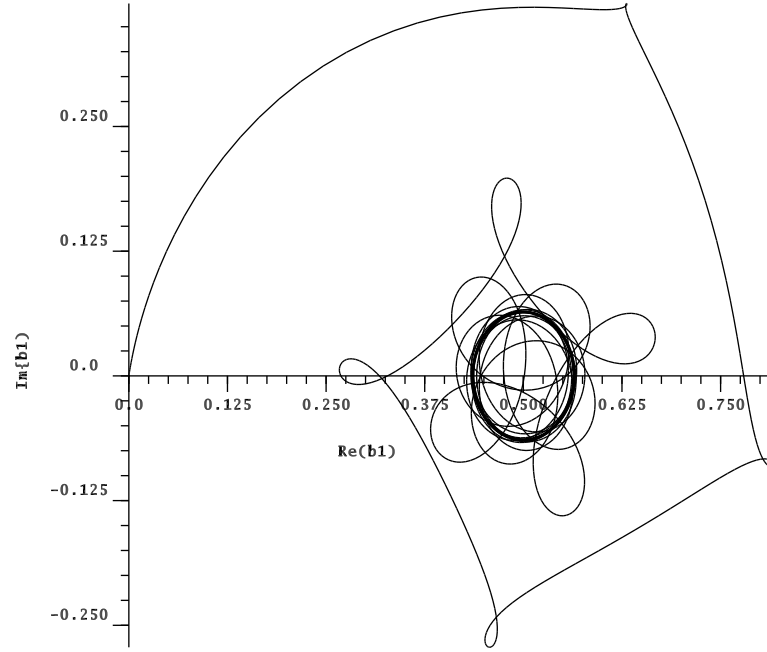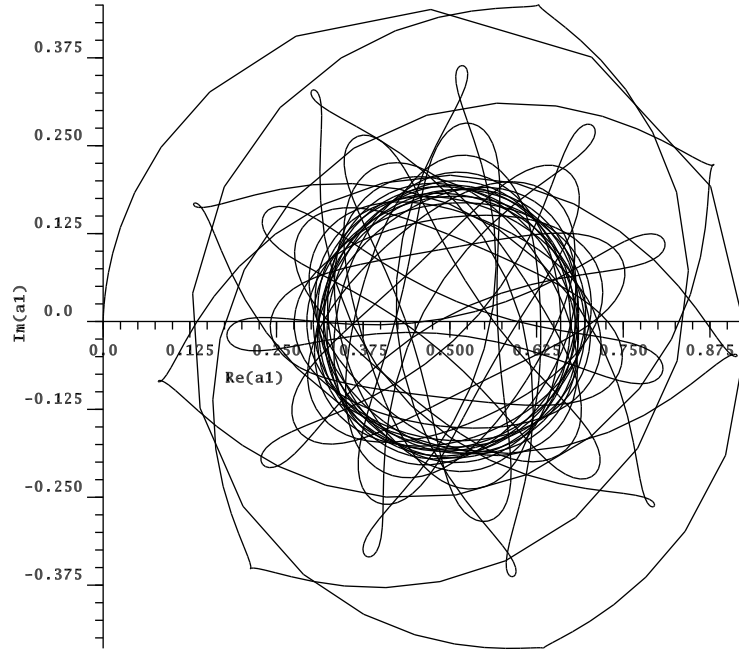
for a larger order.

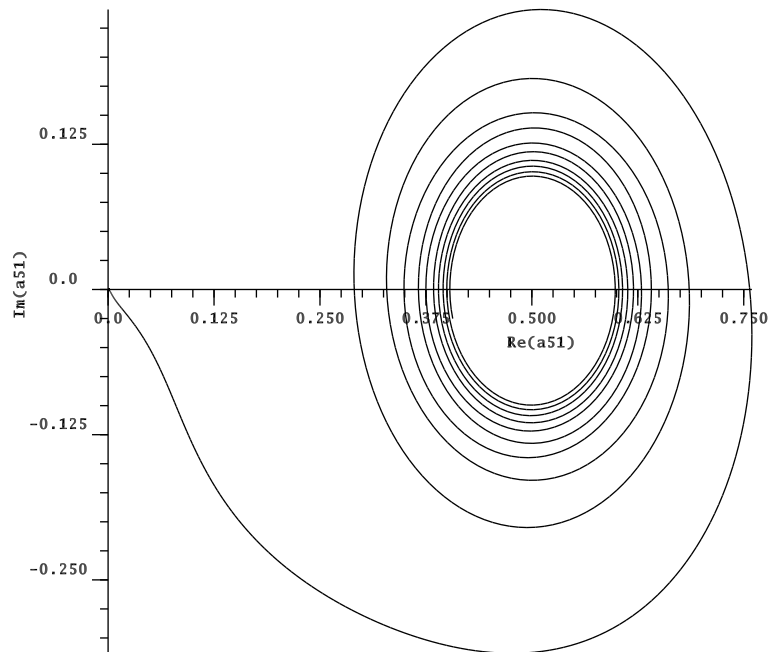Figure 6: The real and imaginary parts of a_1 for a refractive index 2.2-0.022i

Figure 7: The real and imaginary parts of a_{51} for a refractive index 1.29-0.0047i

For large values of x, with a refractive index 1.29-0.0427i, a_{100} becomes close to $(0.5,0)$, but then spirals out to trace circles of radius 0.06 about $(0.5,0)$ for x values ranging from 300 to 500. The quantities a_{200}, a_{300}, a_{400}, behave similarly with circles of .056, 0.055, 0.0546, in general, the an and b_n approach a circle of convergence at about $x = 2n$. What does all this mean? For large x we end up with a series that for (possibly very large) N with terms of similar amplitudes. That is, it is not the calculation of the Mie terms themselves which is the problem, but heavy cancellation in the summation of the series. In the next section we shall investigate the behaviour for large x further.

For small x the a_n and b_n are close to zero and the the singularity in the y_n means that the series can be truncated at smaller values of N. Of course, as $x \to 0$ in the long wavelength limit we get closer and closer to Rayleigh scattering.

Before going on to the large x limit, we now return to the problem of the A_n. Recall that the numerator and denominator vanish for large order because of the $(z/\nu)^\nu$ behaviour in

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eqn.13. We also saw that for large n the A_n ratio is well behaved if the use of the asymptotic expansion made for large n, and that Diermendjian had not noticed how the asymptotic limit could be used. Deirmendjian used recurrence relations for the A_n [3], and Wiscombe discusses more sophisticated and robust recurrence relations developed over some time [2]. We observe two points, the A_n grow almost linearly, and they are multiplied in the denominator by Bessel functions which also have a $(z/\nu)^\nu$ behaviour and vanish for large order. In other words, for practical purposes, the A_n in the numerator are multiplied by zero under the circumstances where the recurrence relations have to be used rather than a straightforward use of Bessel function routines, or indeed the asymptotic relations seen here in eqn.14. As far as the author can tell, all the effort put into the recurrence relations are unnecessary. If there is a small error in A_n at any stage using the asymptotic limit, the a_n will be zero, or effectively zero anyway. One might wonder if the a_n's denominator vanishes along with the numerator for large order, however, the asymptotic relation [4] ensures that this most certainly does not happen, i.e. for fixed z in the limit of large ν,

$$Y_\nu(z) \sim \sqrt{\frac{2}{\pi\nu}} \left(\frac{ez}{2\nu}\right)^{-\nu}. \tag{21}$$

6   More on the Large x Limit

We suppose now, we do not have scattering from a single sphere as discussed in [1], but scattering from a large number of spheres distributed somehow in space. This may approximate for instance a water droplet cloud. It is supposed that the spheres are generally far enough apart so that each sphere is far enough from the others so that the far field approximation holds for the interaction between the spheres. What is more, as is of interest in cloud and atmospheric aerosols, we have a large range of sizes of these spheres. What are the optical properties of such clouds and hazes according to electromagnetic theory.

This is the subject discussed extensively by Diermendjian [3]. Typically, the size distributions are well approximated by log-normal distributions or by modified gamma distributions. These distributions can have long 'tails' into large particle sizes, and so when integrating any quantities over the size distribution to get say, an equivalent extinction coefficient, we may well need to calculate all the Mie theory quantities for large values of x. This is why we took a look at some of the asymptotic relations in the large x limit. (We have some inkling on what will happen from the diagrams in the previous section,)

To add to them, for arbitrary ν, the asymptotic limits of $J_\nu(z)$ and $Y_\nu(z)$ (in the large $|z|$ sense) are [4].

$$J_\nu(z) = \sqrt{\frac{2}{\pi z}} \left\{\cos\left(z - \frac{1}{2}\pi\nu - \frac{\pi}{4}\right) + \exp(|Im(z)|)O(|z|^{-1})\right\}$$

and

$$Y_\nu(z) = \sqrt{\frac{2}{\pi z}} \left\{ \sin\left(z - \frac{1}{2}\pi\nu - \frac{\pi}{4}\right) + \exp(|Im(z)|)O(|z|^{-1}) \right\}. \tag{22}$$

Looking at the formula for the a_n, we see that we have pure real arguments in the numerator, so that the exponential term is small for large x and that explains the oscillatory nature of the a_n and b_n seen in the figures above. For real arguments, the 'envelope' enclosing the oscillations is given by the envelope of the above asymptotic formula, but it is not until the argument is very large, especially for high order, that the phase of the oscillations in the asymptotic expansions matches that of the actual Bessel functions.

So, let us take a look at the A_n, using eqn.21, it is easily found that provided $\kappa > 0$, then

$$\frac{J_{n-1/2}(y)}{J_{n+1/2}(y)} \to i \tag{23}$$

for sufficiently large x. This means that for large x

$$a_n \to \frac{i/m J_{n+1/2}(x) - J_{n-1/2}(x)}{i/m(J_{n+1/2}(x) - Y_{n+1/2}(x)) - (J_{n-1/2}(x) - Y_{n-1/2}(x))}. \tag{24}$$

Again, using eqn.21, one arrives at

$$a_n \to \left(\frac{m}{m+1}\cos p - \frac{i}{m+1}\sin p\right)\exp(ip). \tag{25}$$

Here we have put

$$p = x - (n - 1/2)\frac{\pi}{2} - \frac{\pi}{4}. \tag{26}$$

Switching to complex notation this can be rewritten as

$$a_n \to \frac{1}{2}\left(1 + \frac{m-1}{m+1}\exp(2ip)\right). \tag{27}$$

That is, we expect that for large enough x, the a_n will circle the point in the complex plane $(1/2,0)$ at a rate such that a complete cycle takes place as x varies from X to $X+\pi$. Neither the rate at which a_n rotates, or the diameter of this circle are affected by the order for large enough x, and once x is large enough, the diameter does not decay. This of course requires $\kappa \neq 0$, and so this formula is incorrect for pure real refractive indices. The reason is that for a pure real refractive index the A_n are pure real, so the $A_n \to i$ only applies for a refractive index with a non zero imaginary component. The smaller the imaginary component, the larger the value of x where the A_n approaches i, and for a pure real refractive index A_n *never* approaches i at all. As Deirmendjian points out, for a pure real refractive index the centre of the circle is same as for the complex case, and the diameter is one half regardless of the value of the refractive index or order. At any rate, for a size distribution with a very large slowly decaying tail into large x values we end up with lots of oscillatory terms that do not decay.

In general, there may be numerical problems when adding large numbers of oscillatory terms with similar magnitudes, for instance, in the Fourier series for a Dirac comb function

(a periodic function consisting of Dirac delta functions placed at regular intervals). If we try and perform the (impossible) numerical summation very large narrow spikes soon rise up in the right places, but everywhere else all that is seen is numerical noise. The same is true if we replace the delta functions with say, large narrow Gaussian curves. Here the coefficients do decay, but very very slowly (disregarding for the sake of general discussion the question of exactly how large is large?) So, we are in similar numerical trouble here, and can such problems be remedied?

7   Calculating the Scattering Function

In eqns. 121, 122, 131, and eqn 142 of [1] we encountered the angular distribution of the scattered energy and the phase function. We note that first off, in eqns 121 and 131 we have the product of two infinite series with coefficients a_n and b_n. The functions π_n and τ_n are functions of angle alone. Now, we could integrate over the size distributions over a discrete set of angles and see what the *scattering* function (or*scattering* matrix) actually looks like, but in the problem of multiple scattering in a radiative transfer context, we must have the *phase* function represented in terms of Legendre Polynomials.

The reason for the emphasis is that there can be some confusion. The phase function is normalised. It would be completely wrong to integrate the Legendre coefficients of the phase function over the size distribution as large particles would be under-represented by far.

The scattered energy is proportional to $K_{sc}(x)x^2$. So, we must calculate the Legendre coefficients of the scattering function and integrate these over the distribution. If $L_k(x)$ are the Legendre coefficients for the scattering matrix and $\Lambda_k(x)$ are the Legendre coefficients for the phase matrix, then

$$\Lambda_k = \frac{4 \int_0^\infty \rho(x) L_k(x) \, dx}{\int_0^\infty \rho(x) x^2 K_{sc}(x) \, dx}. \tag{28}$$

To add to the confusion, in [5], Dave defines the matrix elements such as M_1 as the elements of the scattering matrix, and gives the Legendre coefficients L as the Legendre coefficients of the scattering matrix. However in [6] Dave uses the same notation M_1 notation for the phase matrix.

Of course, the infinite series for the scattering functions must be truncated at some value N_{max} depending on the value of the size parameter x. If the particle size distribution has a long slowly decaying tail into large values of x, then the value of N_{max} might be very large indeed.

We could multiply out the series, and integrate the quantities such as $a_i(x) \times b_j(x)$ and obtain the Legendre coefficients by integration over angle at the end of the integration over size distribution. However [5], J. V. Dave has shown that the Legendre coefficients may

be calculated directly from the Mie a_n and b_n coefficients without any angular numerical integration over angles at all. For the convenience of the reader, we repeat the equations later as they arise. The relations between the Legendre coefficients for the scattering function and the Mie coefficients seem to be a rather fearsome summation, however only the inner sums involving the CC^* and like terms are functions of x.

For each x, we must calculate all the a_n and b_n. From these the C's and D's of eqn.20 in [?] can be calculated.

$$C_k = \frac{1}{k}(2k-1)(k-1)b_{k-1} + (2k-1)\sum_{i=1}^{\infty}\left\{\left[\frac{1}{p} + \frac{1}{p+1}\right]a_p - \left[\frac{1}{p+1} + \frac{1}{p+2}\right]b_{p+1}\right\}$$

and

$$D_k = \frac{1}{k}(2k-1)(k-1)a_{k-1} + (2k-1)\sum_{i=1}^{\infty}\left\{\left[\frac{1}{p} + \frac{1}{p+1}\right]b_p - \left[\frac{1}{p+1} + \frac{1}{p+2}\right]a_{p+1}\right\}, \quad (29)$$

where $p = k + 2i - 2$ for the equations for the C and D vectors.

Once this is done, for a single x value, the Legendre coefficients are given by

$$L_k^1 = (k - 1/2)\sum_{m=k'}^{\infty} A_m^{k-1}\sum_{i=0}^{k'} B_i^{k-1}\Delta ik \times Re(D_pD_q^*)$$

$$L_k^2 = (k - 1/2)\sum_{m=k'}^{\infty} A_m^{k-1}\sum_{i=0}^{k'} B_i^{k-1}\Delta ik \times Re(C_pC_q^*)$$

$$L_k^3 = (k/2 - 1/4)\sum_{m=k'}^{\infty} A_m^{k-1}\sum_{i=0}^{k'} B_i^{k-1}\Delta ik \times Re(C_pD_q^* + C_p^*D_q)$$

and

$$L_k^4 = (k/2 - 1/4)\sum_{m=k'}^{\infty} A_m^{k-1}\sum_{i=0}^{k'} B_i^{k-1}\Delta ik \times Im(C_pD_q^* - C_p^*D_q) \quad (30)$$

Note that $k' = (k-1)/2$ for odd k and $(k-2)/2$, for even k, $q = m+i+1+\delta$ where $\delta = 0$ for odd values of k and is unity for even k. Also for odd k, Δ is 1 if $i = 0$ and 2 otherwise, and for even k $\Delta = 2$. Also, $p = m - i + 1$. The reader may note that the terms such as $Re(C_pC_q^*)$ can be written as an upper triangular matrix. Each matrix element will now be integrated over a size distribution, and the Legendre coefficients may be calculated after the integration has been performed.

The L^1 and L^2 are the coefficients for Diermendjian's i_2 and i_1 ([3] eqn.76). In Diermendjian's notation, i_4 has the opposite sign to Dave's D_{21} ([6] eqn.6).

We shall examine a few graphs for the functions C and D to get some feel as to how they behave. First off, we shall just look at the real and imaginary parts of some Cs with a refractive index of 1.3-0.002i as a function of size parameter, and for different series
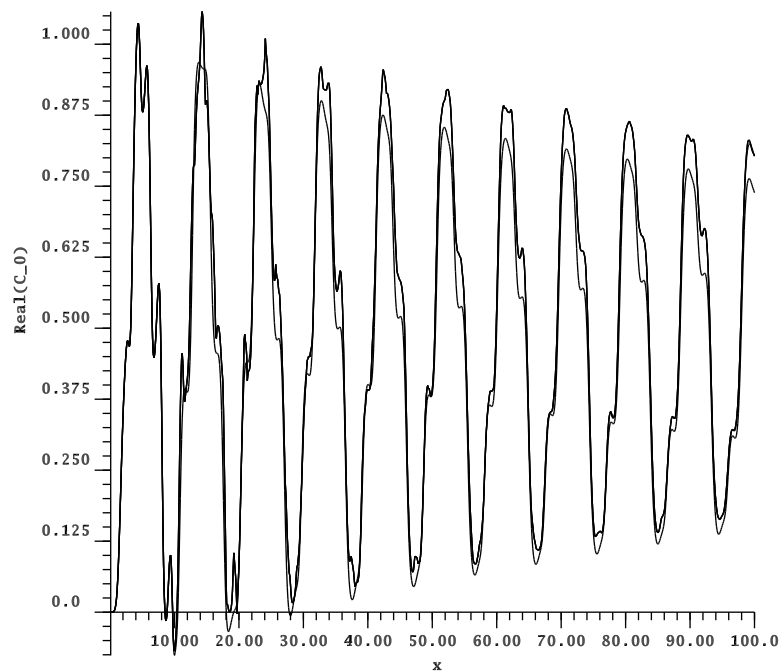
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truncation values.

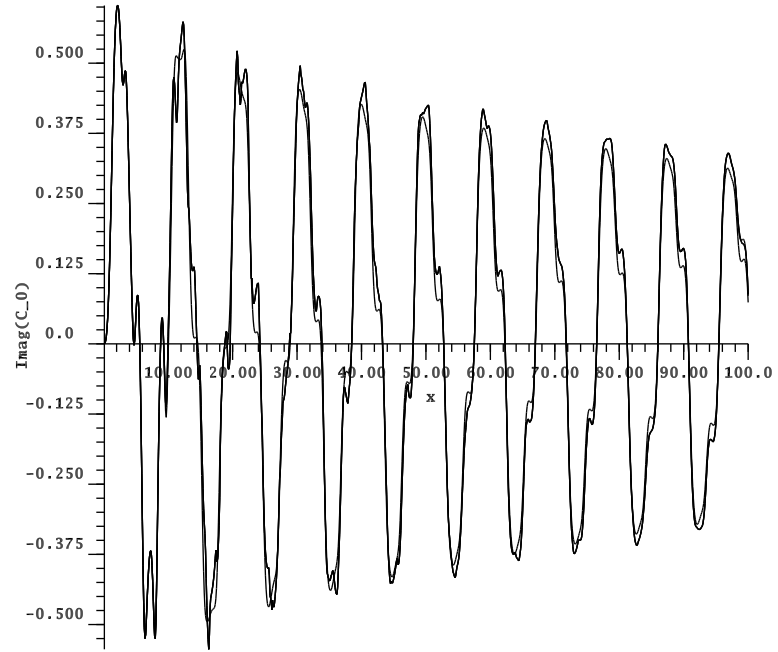Figure 8: A plot of the real part of C_0 for a refractive index of 1.3-0.002i.

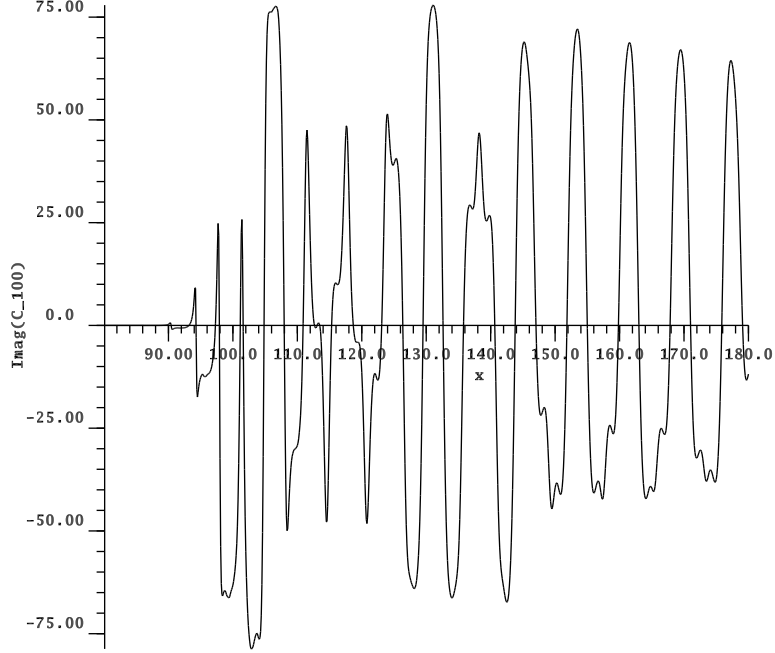Figure 9: A plot of the real part of C_0 for a refractive index of 1.3-0.002i.

Figure 10: A plot of the imaginary part of C_{100} for a refractive index of 1.3-0.002i.

In Fig.8, we plot $Re(C_0(x))$ truncating the series at $i = 10$ and $i = 100$. Truncating at $i = 1000$ alters the values of C_0 by about a tenth of 1% at $x = 100$, and there is no change if the series is truncated at 10 000. At the resolution of the plot there appear to be spikes here and there. However, increasing the resolution shows these to be well behaved smooth curves. So it can be concluded that these are not the signature of numerical noise as discussed earlier at the end of §6. Fig.9 shows the imaginary part of C_0, which oscillates around zero. As n is increased, the C_n behave in similar ways to C_0, except that for x much less than n, the values of C_n are close to zero. For instance, in Fig.10, the real part of C_{100} is close to zero for $x < 90$, the same is true for the imaginary part. As for the D's, the real and imaginary parts display similar features to those of the Cs.

If the real part of the refractive index is increased, while the imaginary part is kept small the small scale features are amplified, and the rate of oscillation increases. In general, increasing the imaginary part smooths out the small scale features. So, we can conclude that the "ripple-spike" problem does not appear to occur when calculating the C and D

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values of eqn.28. However there are small scale features which though they are smooth and well behaved when looked at closely, may well have accuracy problems associated with them when integrating over the size distribution.

To proceed, the integration scheme used by the author is as follows. We step along the x axis in intervals of Δ. The author finds that $\Delta = 1$ is a reasonable value in practise. For each interval, we have an N_g point Gauss-Legendre integration rule, and for each x we calculate the C and D vectors, the upper triangular matrices of products having been initialised at zero are added to after being multiplied by the weights. For instance

$$\int_0^\infty \rho(x) Re(C_p C_q^*) dx =$$

$$\int_0^\Delta \rho(x) Re(C_p C_q^*) dx + \int_\Delta^{2\Delta} \rho(x) Re(C_p C_q^*) dx + \int_{2\Delta}^{3\Delta} \rho(x) Re(C_p C_q^*) dx + \dots$$

$$\approx \sum_{n=1}^{N_g} w_n \rho(x_n) Re(C_p(x_n) * C_q^*(x_n)) + \sum_{n=1}^{N_g} w_n \rho(x_n + \Delta) Re(C_p(x_n + \Delta) * C_q^*(x_n + \Delta))$$

$$+ \sum_{n=1}^{N_g} w_n \rho(x_n + 2 * \Delta) Re(C_p(x_n + 2 * \Delta) * C_q^*(x_n + 2\Delta)) + \dots, \tag{31}$$

where the x_n are the quadrature points on the interval $(0, \Delta)$ (in practise this will be the interval (0,1) as stated. We stop when the size distribution function becomes very small. So there are two factors controlling the accuracy apart from where to truncate the Mie series. That is, the decision of how many intervals are needed (or how far to march along the x axis), and the order of quadrature rule to use. The upper triangular matrices should, of course, be stored in compact vector form.

Now we return to eqn.29. We shall repeat Dave here [5], hopefully we shall make it easier to follow than in the original. The start value of the outer sum over m is 0 for $k = 1$ and $k = 2$. It is 1 for $k = 3$ and $k = 4$, and it is 2 for $k = 5$ and $k = 6$. The sequence is obvious. Now if m is greater than the starting value for the series, that is $m > k'$, and if k is odd

$$A_m^{k-1} = \frac{(2m - k)(2m + k - 1)}{(2m + k)(2m - k + 1)} A_{m-1}^{k-1}. \tag{32}$$

If k is even, then

$$A_m^{k-1} = \frac{(2m - k + 1)(2m + k)}{(2m + k + 1)(2m - k + 2)} A_{m-1}^{k-1}. \tag{33}$$

Of course, we need the initial values for $m = k'$. For $k = 1$ $A_0^0 = 2$, and for $k = 2$ it is $A_0^1 = 4/3$.

If k is larger than 2, the above recurrence relations still hold, however there is another recurrence relation for the starting value $m = k'$. Again, it depends on whether k is odd or even. If $k = 3, 5, 7, \dots$ then the start value is

$$A_{(k-1)/2}^{k-1} = \frac{4(k-1)(k-2)}{(2k-1)(2k-3)} A_{(k-3)/2}^{k-3}, \tag{34}$$

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and for $k = 4, 5, 8, \ldots$

$$A^{k-1}_{(k-2)/2} = \frac{4(k-1)(k-2)}{(2k-1)(2k-3)} A^{k-3}_{(k-4)/2}. \tag{35}$$

Similarly, if $i > 0$, then for odd k

$$B^{k-1}_i = \frac{(k-2i+1)(k+2i-2)}{(k-2i)(k+2i-1)} B^{k-1}_{i-1}, \tag{36}$$

and for even k

$$B^{k-1}_i = \frac{(2i+k-1)(2i-k)}{(2i-k+1)(2i+k)} B^{k-1}_{i-1}. \tag{37}$$

For $k = 1$ the start value b^0_0 is 1, and for $k = 2$ it is b^1_0 which is $1/2$. After which the start values are

$$B^{k-1}_0 = \frac{(k-1)(k-3)}{k(k-2)} B^{k-3}_0 \tag{38}$$

for even k, and

$$B^{k-1}_0 = \left(\frac{k-2}{k-1}\right)^2 B^{k-3}_0 \tag{39}$$

for odd k.

It is noted that the Ls of eqn.29 are the Legendre coefficients for the scattering functions. To normalise, we must also calculate the scattering efficiency. That is

$$K_s = \frac{2}{x^2} \sum_{n=1}^{\infty} (|a_n|^2 + |b_n||^2). \tag{40}$$

Also of interest is the extinction efficiency

$$K_e = \frac{2}{x^2} \sum_{n=1}^{\infty} Re(a_n + b_n). \tag{41}$$

We introduce

$$\overline{K}_{sc} = \int_0^{\infty} x^2 \rho(x) K_{sc}(x) \, dx$$

and

$$\overline{K}_{ex} = \int_0^{\infty} x^2 \rho(x) K_{ex}(x) \, dx \tag{42}$$

and the single scattering albedo for the distribution

$$\varpi_0 = \frac{K_{sc}}{K_{ex}}. \tag{43}$$

To get from the coefficients for the scattering function to the normalised phase function we must multiply the coefficients by $4/\overline{K}_{sc}$.

8 Size Distribution

The author is most interested in atmospheric applications. Here the modified gamma distribution is a very common model. That is the probability density function for finding a particle with radius between r and $r + dr$ is

$$\rho(r) = ar^{\alpha}e^{-br^{\gamma}}. \tag{44}$$

Setting the derivative to zero gives us the mode radius

$$r_m = \left(\frac{\alpha}{b\gamma}\right)^{1/\gamma}. \tag{45}$$

In practice, a mode radius is given for a particular distribution, from which b is calculated. We use the result (which follows from the definition of the Γ function)

$$\int_0^{\infty} x^{\nu-1}e^{-\mu x}dx = \frac{\Gamma(\nu)}{\mu}, \tag{46}$$

so that normalisation gives us

$$a = \frac{\gamma b^{(\alpha+1)/\gamma}}{\Gamma((\alpha+1)/\gamma)}. \tag{47}$$

The mean radius is then

$$< r\rho(r) >= \frac{a\Gamma((\alpha+2)/\gamma)}{\gamma b^{(\alpha+2)/\gamma}}. \tag{48}$$

The average values of the square and the cube are found by replacing $\alpha + 2$ with $\alpha + 3$ and $\alpha + 4$. These are important if we are given the total mass or volume of the particles in the cloud or haze. As to values of α and γ, Deirmendjian uses the following values for α and γ in his definitions of distribution types. distribution [3].

Type	N	r_m	α	γ
Rain M	$100cm^{-3}$	$0.05\ \mu m$	1	1/2
Haze M	$1000cm^{-3}$	$0.05\ mm$	1	1/2
Haze L	$10cm^{-3}$	$0.07\ \mu m$	2	1/2
Rain L	$1000cm^{-3}$	$0.05\ \mu m$	2	1/2
Haze H	$100cm^{-3}$	$0.1\ \mu m$	2	1
Hail H	$10cm^{-3}$	$4.0\ \mu m$	2	1
Cumulus C.1	$100\mu m^{-3}$	$4.0\ \mu m$	6	1
Corona C.2	$100\mu m^{-3}$	$0.05\ \mu m$	8	3
MOP cloud C.3	$100\mu m^{-3}$	$2.0\ \mu m$	8	3
Corona C.4	$100cm^{-3}$	$4.0\ \mu m$	61	3

In other cases, the log-normal (or lognormal) distribution is a good description. Both the modified gamma and lognormal distribution are used by Shettle and Fenn [7] in their classic work. The standard form for the lognormal distribution is

$$\rho(r) = \frac{e^{-[(lnr-\mu)^2/2\sigma^2]}}{r\sigma\sqrt{2\pi}}. \tag{49}$$

The moments are given by

$$\mu_k = \int_0^\infty r^k \rho(r)dr = e^{\mu k + k^2\sigma^2}, \tag{50}$$

so that the mean radius and mean area and volume are easily calculated.

When the lognormal distribution is used there can be some confusion. Experimenters will use logs in base 10 in their results as this is much easier to understand visually, and therefore tend to use logarithms in base 10 when defining a lognormal fit to data.

As changing bases in logarithms is easily forgotten, we make a brief mention of it here. If

$$y = a^x,$$

then

$$y = (b^{\log_b a})^x,$$

so

$$\log_a y = x,$$

$$\log_b y = x \log_b a,$$

in which case

$$\log_a y = \frac{\log_b y}{\log_b a}. \tag{51}$$

So, if we use logs to the base 10 in the argument of the exponential in eqn.46 and write $\mu = \log_{10} r_m$ we have

$$\rho(r) = \frac{e^{-[(ln(10))^2 (\log_{10} r - \log_{10} r_m)^2 / 2\sigma^2]}}{r\sigma\sqrt{2\pi}}. \tag{52}$$

Now if we write $\overline{\sigma} = \sigma/ln(10)$ we have the form

$$\rho(r) = \frac{e^{-[(\log_{10} r - \log_{10} r_m)^2 / 2\overline{\sigma}^2]}}{ln(10)r\overline{\sigma}\sqrt{2\pi}}. \tag{53}$$

Despite the odd looking mix of logarithms to different bases and the the number e, this is quite a natural way to present experimental data. Indeed, the distribution is used in this form in [7]. All the moments of eqn.47 follow immediately on recalling that $\sigma = ln(10)\overline{\sigma}$. Note that the values of σ tabulated in Shettle and Fenn are $\overline{\sigma}$ according to the notation above.

We note that the scattering cross sections will vary very roughly with the size parameter squared, which means that even when the pdf has small values at large r, the contributions from the particles may still be large.

9 The Computer Code

There are downloadable computer codes at this site for calculating all that has been outlined so far. These can be found via http://seagods.stormpages.com. Here we shall describe how to use them. Note that any dependencies not downloadable elsewhere (in this case Mydbesj) can be found under Mylib.

The programs are called Scalar.cpp, and Vector.cpp. Naturally Scalar.cpp computes the scalar scattering function, while Vector.cpp calculates the four components of the scattering matrix.

If we are given a complex refractive index of for the particles and the wavelength, and a size distribution we can calculate all the properties we need. The size distribution is in the code can be either the modified gamma distribution which is completely described by the values α, β, and a mode radius, or a lognormal distribution as defined by σ and a mode radius.

For the moment we suppose that the code has been downloaded, and the user needs to know how to use it.

Apart from the physical parameters and the size distribution type, the user must supply the order of Gauss quadrature for each subsection for integration over the size distributions. Where to truncate the series, is calculated internally. The user must supply a tolerance for the accuracy of the integral of $x^2\rho(x)$ to be calculated. This gives a measure of how much of the size distribution has been captured. The factor of x^2 is used because we are integrating quantities which grow as x^2, so the integral of $\rho(x)$ might be very close to one, but we shall still be chopping off the tail of the distribution too early if we use this latter integral as a measure. The value of x at which the tolerance is met determines the truncation of the Mie series at $N_{max} = vx_{max} + w$. The recommended values are $v = 2$ and $w = 10$ [5] p1890. However, the user is free to increase either and they are part of the user input.

Unfortunately there is yet another complication. We have found that cutting off when the integrals of $\rho(x)$ and $x\rho(x)$ is not accurate enough because we are integrating quantities that grow as x^2. However, our size distributions are theoretical model distributions. We might find that for certain distributions, for even modest tolerences of 0.0001, we must integrate up to particle sizes of the order of the size of golf balls! So, we are forced to introduce a sensible cut-off to the distribution and renormalise it.

For instance, "dust like" aerosols have a bimodal distribution. The larger of the two modes has a mode radius of $0.5\mu m$, and the distribution is lognormal with $\sigma \approx 0.45$. If we put in a tolerance of 10^{-5} for the mean surface area, we must integrate up particle sizes of $\approx 5mm$. A sensible cut of here might be $0.1mm$. This would not be sensible for a cloud droplet distribution with a mode radius of $10\mu m$.

This seems to introduce some arbitrariness, but whether the cut-off comes from some physical model — or just the users "common sense", it has to be there. If it were just a matter of an aerosol model (say) having the occasional pea or even golf-ball sixed particle in a cubic kilometre of air, we could just shrug and say, well — that's the properties of such a model distribution.

Often this would be fine, especially so when the the wavelengths are long compared the mode radius. The values of x would still be manageable and the series truncation value would remain small, If the opposite is the case, we hit real trouble. It would mean that we would have Legendre series with tens of thousands of terms, and require the entire computing power of the planet to calculate them. Often we may wish to calculate a spectrum of values for the same distribution. It would make little sense to have no cut-off for infrared wavelengths, but a cut-off for ultra-violet.

So if we have a cut-off for one wavelength we must have a cut-off for the all the others. Whatever happens we have a maximum value $nsteps$ which tells us where to truncate the

series. The user must explain what cut-off values were used and why to any user of the data. If course, if we require that there must not be a cut-off, we can input a cut-off so large that it will not be reached.

Of course, this maximum value could be used to truncate the series for all values of x, but we can gain considerable speed up if we use the same rule to $N = vx + w$ for each x. The user has an option as to whether this speed up is used, but the speed up is recommended.

(How these and other parameters are input will be described shortly.)

We recommend that the user experiments as in the following manner. Set a coarse integration rule first, and choose not to calculate the Legendre moments. Two output files, distfun.dat and distfun_xsq.dat enables the user to see how the p.d.f, and the p.d.f times the square of the size parameter x behaves over the interval. The output will include also the numerical approximation to the area under the pdf curve, the numerical approximations to - and analytical values of: the mean radius; mean surface area; and mean volume.

Next, the user can increase the resolution of the integration (while still omitting to calculate the Legendre moments) to get an idea as to how the optical depth per kilometer and the single scattering albedo converge. Then, the user can start to look at how the Legendre moments are behaving. Depending on the distribution function parameters, the mode radius, and wavelength, this can take a considerable amount of computing time. The integration is performed by stepping along the x axis in units of $x = 1$ at a time, and performing a gauss quadrature on each subinterval.

There are two things that can be mentioned at this point. The first normalised phase function moment should be equal to one. So, how close to one the numerical integration gets is at least some indication whether the code is getting there or not. The second moment is 3 times the asymmetry parameter mentioned earlier, and is also an indicator for convergence. *However, the Legendre coefficients output by the code are the L_k of the unnormalised scattering matrix. The quantity $\overline{K}_{sc}$ is part of the output, so the normalisation must be made by the user using the NormaliseS or NormaliseV codes as appropriate.*

The reason for not normalising is that it may be envisaged that the numerical integration may be required to done in 'chunks'. The reason for this is that the calculating the moments can take a very long time time when the depending on the wavelength and the size distribution parameters. So, breaking up the calculation means that we cannot normalise for any particular 'chunk'. The normalisation has to be done after using one of the 'Stitch' programs to join the chunks together.

Splitting up the calculation will be necessary if the computer on which the code is run has to be switched off overnight, and the integration takes more time than the 'on time' of the

computer. Of course, if the user has access to many machines over a network, or has parallel processing capability, the user can split the problem up to reduce run-time. Splitting the integration into parts and adding all them all up at at the end is then the way to proceed.

This is the reason that the Legendre coefficients of the scattering matrix are output rather than the Legendre coefficients of the normalised phase function. It is also up to the user to rename all the output files (in a script or manually). So, the output files from the "Vector" program should be renamed "Vector_part1.dat", "Vector_part2.dat", and so on. If the scalar program is split up, the outputs should be renamed "Scalar_part1.dat", and so on. This renaming could be done manually, or in a script written by the user.

There is further code to add up all the sub integrations for the $\overline{K}_{sc}$, $\overline{K}_{ext}$, and the $\overline{L}_k$ so as to calculate the Λ_k for the distribution, the extinction coefficient per metre for the distribiution, and the single scattering albedo. These are StitchS.cpp and StitchV.cpp which expect the input files described above. To convert to normalised Legendre coefficients, use NormaliseS or NormaliseV after the Stitch programmes have been run.

9.1   input

At this point we can introduce the input file which must be named "input.dat". An example is

```
1.3  0.02    0.5e-6
2.0  0.5    2e-6   100.0
16 1e-6   0.1
1 1 1
2.0   10
0 100 200
```

The first line consists of n_r, κ, and λ. The complex refractive index in the code is $m = n_r - i\kappa$, so don't forget that κ has to be positive in the input file. Also, note that the wavelength is in metres, so that the input wavelength is in the visible part of the spectrum.

The second line are the two parameters of the size distribution, and the next number is the mode radius (again in metres). The last is the number density. Much data are in the numbers per cc, so in the input file, the number *must have* units of per cubic centimetre. The two parameters for the size distribution are α and γ for the modified gamma distribution. If the lognormal is selected (see line 4 below) the first parameter is $\overline{\sigma}$, *and the second parameter is a dummy but must be there.* Another thing to recall is the the value of σ should actually be the value according to the transformed lognormal of eqn.50,

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that is enter $\overline{\sigma}$ rather than σ.

The third line consists of three numbers. The first is the order of gauss quadrature for integration over each subinterval of the size distribution. The second number is the tolerance as discussed in the previous section. Decreasing this will increase the cutoff in x and increase the the maximum truncation value for the size distribution (called *istop* in the actual code). Lastly, there is the value of the cut-off in the size distribution. This is entered in millimetres. I have chosen this strange discrepency in units because if there is no *physical* reason for the cut-off, we have an easy feel for the size in everyday life.

The fourth line consists of two boolean variables, and an option number. The first boolean variable is 1 if we truncate the Mie at smaller values where x is small in the integration. If not, it will truncate at the value of *istop* whatever the integration point. The second boolean should be set to 1 if the Legendre moments are to be calculated, and to 0 if only the extinction coefficient and single scattering albedo are required. The option number is to be set to 1 if the modified gamma distribution is used and two if the lognormal is to be used. (In fact, it will be the lognormal if any number other than one is used.)

The fifth line tells the how to calculate the cutoff if the low order truncation (line 4) is used for small x. These numbers should be integers, and the values of 2 and 10 are recommended as stated above.

The last line consists of a bool which tells the code whether the the integration is to be split up into chunks or not. If it is not to be split up, the value of *istep* shall run from zero to *nsteps*, which is where the tail of the distribution is cut off according to the value of the tolerance. Nonetheless, two integers are required next, and these shall be overwriten with zero and *nsteps* internally.

If the integration is to be split up, then *istep* will run from the first integer (*chstart*) up to the second integer (*chstop*). The value of *chstop* for the last chunk where the tail of the distribution should be *nsteps*. The user will know what this value is from experimenting with the tolerance, which outputs the value of *nsteps* to the screen. (Note both these integers need to be there if the calculation is *not* to split up in this way. They will be ignored. Set them to something like "-1000" so when you look at the input files you know the numbers mean nothing.

Please note that *there are no safety belts*, that is the program will just crash and burn if there is any inconsistency in the input, or worse still produce believable looking garbage. It is entirely up to the user to understand the input and get it right, or when writing a script to call the program, to get their own script right for all circumstances.

9.2 output

The output file is "Scalar.dat". It includes some of the input data so as to keep track of which file is which. (For multiple runs the author is used to writing shell-scripts to create each input file and copy the output file to a different name after the run.)

So, the first line of output is the mode radius, and α and β of the size distribution. The second line consists of the n_r and κ.

The first number on the third line is the optical depth for metre at the given number density per cc. The second is the single scattering albedo, the third is $\overline{K}_{sc}$, the fourth is $\overline{K}_{ex}$. and the fifth is the number density. *The number density is in numbers per cc, even though it's MKS in the program.* The next line consists of the mean radius, spherical surface area, and volume. This enables the surface area and volume of scatterers per cubic metre for the input number density. The next line consists of four integers, *nsteps, istop, chstart* and *chstop*. The values of *chstart* and *chstop* will be zero and *nsteps* if the integration has not been split into several or even many runs. (Recall that *istop* is the truncation value for the Mie series and so gives the number of moments in the output.)

The following lines are the Legendre moments of the scattering matrix or scattering function in the scalar case. These are in order, one per line or four per line for the vector case. If there are many files, one per chunk, then these numbers must all be added to make one final output file.

Once we have the single output file, we can find the moments of the normalised phase function. We simply read in the output file and do the normalisation described above and write it to a new file.

If the first normalised moment is close enough to one, but not quite the user *could* normalise them by dividing all the moments by the first. However, this is a sign of poor convergence or of chopping off the distribution tail too early. So it is best to calculate again with a finer Gauss quadrature rule in eqn.31, and/or a lower tolerance.

10   A Few Numerical Experiments

We start off with a particularly challenging distribution, the large component of the Shettle and Fenn dust aerosols [7]. This has a value of $\overline{\sigma}$ of 0.475, and a mode radius of $0.5\mu m$. At a wavelength of $0.337\mu m$ this has a refractive index of 1.53-0.008 i as obtainable from HITRAN [8].

A few years back, the author had a 1GHz PC with a 0.5Gbyte RAM and a Linux operating system. So the reader will now see how the code performed under these fairly limited circumstances.

First, we computed the components of the scattering matrix rather than just the scalar phase function. With a tolerance of 1.0e-3, the number of unit intervals for Gauss quadrature is 2996. If we set *chstart* to 2995, *chstop* to 2996.and use a 16 point Gauss quadrature over this interval, the C clock function gave a time of 334 seconds for the integration over the unit interval. However, in reality, the time duration was much longer than this. There was heavy paging going on in virtual memory, the C clock function just doesn't 'tick' when paging. The real time duration was more like half an hour. Also, the computer slowed down to a crawl for any other functionality, even though the priority was set to a minumum using the "nice" command. In effect, the machine was completely tied up. If the programme was run without minimising the priority, the machine was "frozen" completely.

The run time for the calculation was simply too much, however, if we reduced the tolerance, so that the integral of x^2 times the pdf was short by 0.25% rather than 0.1%, the integration time per unit interval is reduced to about three to four minutes, and the machine did not freeze. This then required integration over 2198 unit intervals. The reduction in required RAM is just enough for the calculation to be made at all. Of course, these days it is easy to have get a machine with much more RAM and a much faster clock rate as well as four processors.

Although we have given a rough guide, we include this section so the reader can see how things behave in practice. First we start off with a modified gamma distribution with $\alpha = 1$ and $\gamma = 1/2$. The mode radius is set at $r_m = 0.05\mu m$ which corresponds to Deirmendjian's Haze-M model. We use a refractive index of $1.34 - 0i$ at a wavelength of $0.45\mu m$. This corresponds to Table T.1 of [3]. An 8 point quadrature rule was used on each sub-interval.

The distribution function multiplied by x^2 is plotted below. It seems as though we
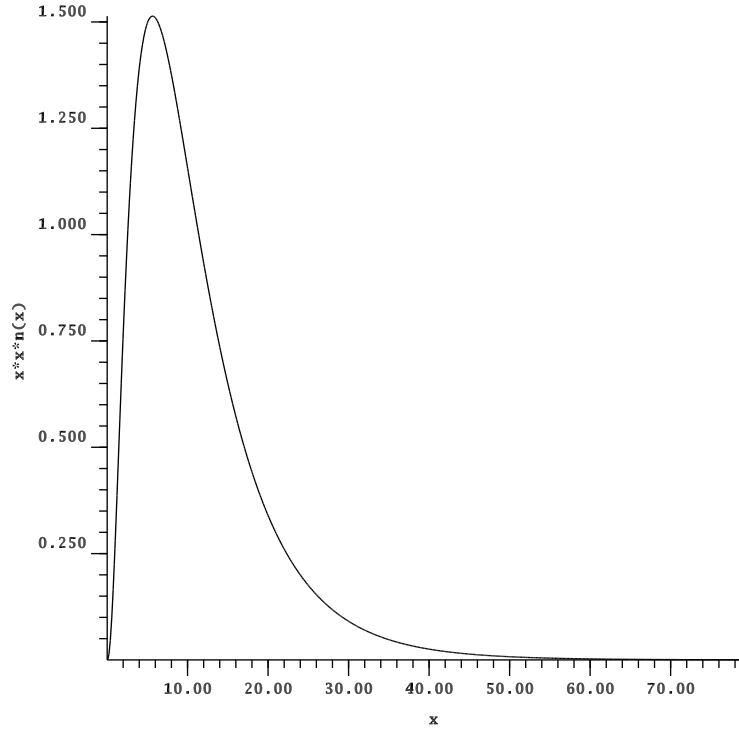
shouldn't bother about any $x > 50$,

Figure 11: The modified gamma distribution $\times x^2$

We first supposed that we could truncate the series after the 50th term, however this produces spurious oscillations and "looks wrong". Truncating at $x = 100$ produces something very close to the phase function that is required though. Only the third significant figures change on increasing the cutoff, and going to higher order quadratures produces little change.

The phase function corresponding the above values is plotted below. The forward
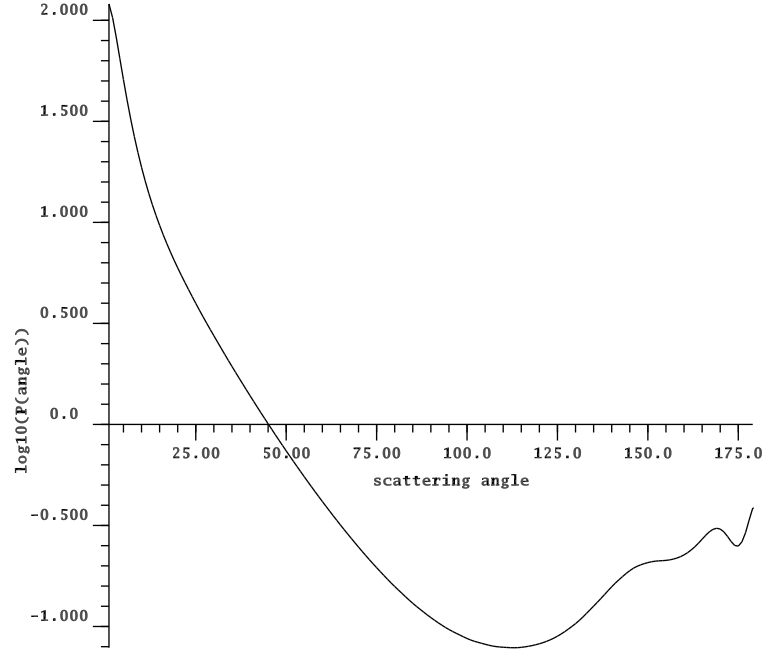

Figure 12: The Phase Function for Haze-M at $\lambda = 0.45\mu m$

scattering "spike" is typical, as are the features in the backscattering half of the graph. At any rate, the resulting phase function is differs from the tabulated values in T.1 only in the third significant figure.

So, as a rough guide, figure out the value of x so that x^2 times the pdf is no longer significant, double it, add 10 to it, and use that as the truncation value, and you won't be far off. In the above example, stopping at $x = 60$ and truncating after 130 terms gives a runtime of just under 3 seconds on a 32 bit 800MHz PC (with an 8 point Gauss quadrature). Earlier, we mentioned a run-time of 45 minutes. With more extreme size distributions this can be exceeded very easily. Especially where RAM is limited so that the compiler starts paging in virtual memory. Depending on the wavelength and size distribution parameters, the calculation become very demanding on computing power.

We find that a fairly high tolerance of 0.001, and an 8 point quadrature reproduce Diermendjian's results. Decreasing the tolerance to $1e - 7$ and increasing the quadrature

rule to 32 give much more accurate results. (Diermendjian simply did not have the processing power.)

Earlier, we mentioned two ways of calculating the moments by integration over angle. The author has not tested all ways to compare them. Comparison is dependent on the amount of RAM, the machine architecture, whether parallel processing is available, and so on. Not only that, but a true comparison would require each method to be programmed to maximum efficiency (again depending on the machine). The author has gone for Dave's method, and has no inclination to code up all three methods.

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