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REPAD: An empirical model of pitch angle distributions for energetic electrons in the Earth's outer radiation belt

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REPAD: An empirical model of pitch angle distributions for energetic electrons in the Earth's outer radiation belt

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Key Points:
- An empirical model for pitch-angle distributions of outer-belt e- from data
- Model is validated and evaluated, using RBSP data, and can be applied to GPS
- The e- PADs are highly stable for L ~4-5 but not outward (e.g., GEO with L ~6.5)

Supporting Information:
- Readme
- Figure S1a
- Figure S1b
- Figure S2a
- Figure S2b
- Figure S3
- Figure S4
- Figure S5
- Figure S6
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- Figure S8a
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- Figure S10
- Text S1

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Citation:

Abstract
We have recently conducted a statistical survey on pitch angle distributions of energetic electrons trapped in the Earth's outer radiation belt, and a new empirical model was developed based upon survey results. This model—relativistic electron pitch angle distribution (REPAD)—aims to present statistical pictures of electron equatorial pitch angle distributions, instead of the absolute flux levels, as a function of energy, L shell, magnetic local time, and magnetic activity. To quantify and facilitate this statistical survey, we use Legendre polynomials to fit long-term in situ directional fluxes observed near the magnetic equator from three missions: CRRES, Polar, and LANL-97A. As the first of this kind of model, REPAD covers the whole outer belt region, providing not only the mean and median pitch angle distributions in the area but also error estimates of the average distributions. Preliminary verification and validation results demonstrate the reliable performance of this model. Usage of REPAD is mainly to predict the full pitch angle distribution of fluxes along a given magnetic field line, or even on a given drift shell, based upon one single unidirectional or omnidirectional flux measurement anywhere on that field line. This can be particularly useful for data assimilation, which usually has large tolerance on data errors. In addition, relatively small variations in pitch angle distributions measured at L shell between ~4 and 5 justify the assumption of fixed pitch angle distributions at GPS equatorial crossings (L ~4.2) used in our previous studies.

1. Introduction

Studying the pitch angle distribution (PAD) of trapped radiation belt particles has attracted intensive interests since the early days of magnetosphere research. This is mainly driven by the academic curiosity of understanding the cause of observed and varying PADs, which are often related to processes governing the dynamics of radiation belts. For example, energetic electrons (~MeV) observed at geosynchronous orbit experience strong enhancements after some geomagnetic storms but significant losses after others [Reeves et al., 2003; Chen et al., 2007b]. This should reflect the competition between various processes in each individual event, and tracking distributions and variations of electron PADs may help identify and evaluate those processes.

Meanwhile, practical needs also demand knowledge on electron PADs. For example, although the most efficient way to monitor radiation belts is to make fully directional measurements close to the magnetic equator, satellite orbits are often not so ideal due to other considerations and particle instruments on board may only measure electrons in either a single direction or omnidirection, such as GPS satellites in the latter case. Therefore, to maximize the usage of observations with limited resolution on pitch angles, it will be really useful to know not only the statistically averaged PAD but also its reliability. This work addresses this issue by developing a new PAD empirical model.

There exist several typical PAD types for electrons. The simplest one is the isotropic distribution, which is often observed in newly injected substorm electrons in the midnight sector at geosynchronous (GEO) orbit [Åsnes et al., 2005]. Other well-known PADs include the pancake distribution, with a peak flux at 90° pitch angle, and the butterfly distribution, with a local minimum at 90° and a maximum before reaching each loss cone. The latter is also referred to as a cigar distribution sometimes. For relativistic electrons observed at GEO, pancake PADs generally dominate the dayside, while butterfly the nightside [West et al., 1973; Baker et al., 1978], and this difference is mostly due to the drift-shell splitting of electrons in a nonsymmetric magnetic
Another common PAD is the flattop distribution, characterized by a near-flat distribution over a broad range of pitch angles centered at 90°. A previous study by Horne et al. [2003] has argued that the flattop PAD is the signature of in situ acceleration caused by wave-particle resonance, which can be differentiated from the pancake distribution resulting from inward radial diffusion. One example for each typical PAD is shown in Figure 1.

Electron PADs are not a new research topic. For instance, empirical radiation belt models, such as AE-8, provide average flux levels for the whole radiation belt region, which may implicitly provide PADs. However, since data samples used for the model are neither distributed evenly along each magnetic field line nor acquired simultaneously, the derived PADs are inevitably biased to some unknown degree. In an attempt to update outer zone energetic electron environment, Vampola [1998] conducted a statistical survey on CRRES electron data and presented the pitch angle distribution coefficient as a function of L shell, with the exclusion of butterfly distributions due to the form of selected fitting function. In addition, there are also other case and statistic studies directly focusing on electron PADs [e.g., Horne et al., 2003; Gannon et al., 2007; Gu et al., 2011] that establish connections between variations in electron PADs and factors such as wave activities and storm events. Although revealing interesting physics, results from those studies are very useful for interpreting pitch angle resolved observations but not for non–pitch angle resolved observations.

In this report, we present a new empirical model for outer belt electrons. Based upon a statistical survey on electron observations accumulated during the last two decades, this relativistic electron pitch angle distribution (REPAD) model provides reliable PADs with error bars for energetic electrons along field lines trapped within L shells between three and nine. In addition, we reintroduce an effective PAD-fitting method that will be a very useful tool for future mining through large volume of pitch angle resolved data, such as data from the Van Allen Probes mission.

Descriptions on instruments, data, and preprocessing are given in section 2. Section 3 describes the fitting scheme, its efficiency, and presents model parameters. Some model results are presented in section 4, and section 5 shows model verification, validation, and evaluation results. Section 6 discusses the sample application to GPS data, and this report is concluded by a summary in section 7.

2. Instruments and Data

Long-term electron observations used here are from the Combined Release and Radiation Effects Satellite (CRRES, operating 1990–1991), Global Geospace Science Polar Satellite (Polar, 1996–2008), and one Los Alamos National Laboratory geosynchronous satellite LANL-97A (1999–2008). These three satellites have very different orbit configurations as described below, and the whole data set expands over ~22 satellite years in total.

CRRES was launched in July 1990 into a low-inclination (~18°) geosynchronous transfer orbit (GTO) until the mission was unexpectedly ended due to the loss of communication in October 1991. With the fluxgate magnetometer instrument aboard measuring local magnetic field [Singer et al., 1992], the Medium Electrons A (MEA) [Vampola, 1992] spectrometer on CRRES provides pitch angle resolved electron data covering the energy range from 0.15 up to 1.7 MeV in 17 differential channels. MEA data used here have a time resolution of 1 min. The advantage of CRRES mission is its GTO which has good coverage of both L shells and the full PADs, and the disadvantage is the short mission period and the lack of data in the prenoon sector. Figure 2a presents one MEA data example, showing pitch angle resolved fluxes for 148 keV electrons on 1 day. Butterfly distributions observed at ~0100 UT and pancake distributions at ~0730 UT correspond to ~2200 and 1600 MLT (not shown), respectively, in the outer belt. Due to its high data quality, we will often use MEA data as examples later on.
The Polar satellite has a high-inclination (~86°) polar orbit of 2 × 9\ RE, and the mission started in February 1996 and ended April 2008. Polar crosses the magnetic equatorial plane every ~9 h. Due to orbital precession, Polar’s equatorial crossing keeps changing its radial distance with time slowly (e.g., it was around L ~ 4 in 1996 and L ~ 9 in 2002) and sweeps across all magnetic local times (MLTs) every ~12 months. This orbit configuration for Polar has the best spatial coverage among all three missions. With the Magnetic Field Experiment [Russell et al., 1995] measuring local magnetic field, the comprehensive energetic particle and pitch angle distribution (CEPPAD) experiment [Blake et al., 1995] onboard Polar provides pitch angle resolved flux data of energetic electrons, covering the energy range from 30 keV up to 10 MeV with detectors of the Imaging Electron Sensor and the High Sensitivity Telescope subsystems. Flux data used here have an average time resolution of 24 s. Only data with L > 3 are used to avoid potential contamination from protons [Friedel et al., 2005].

Figure 2. Data samples from three satellite missions. (a) Time series of pitch angle resolved fluxes for 148 keV electrons observed by CRRES on 1 day (20 April 1991) and (b) the corresponding L shell curve (black) and the ratio curve (gray) between local and equatorial magnetic field. Equatorial pitch angle values (black) corresponding to five magnetic ratios are also given to the right of the panel. (c and d) Polar fluxes for 782 keV electrons, L shells, and magnetic field ratios on 30 May 2001. (e and f) LANL-97A fluxes for 1368 keV electrons, L shells, and magnetic field ratios on 11 December 2002. Gray gaps in fluxes are due to the lack of measurements.

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LANL-97A measures electrons in geosynchronous (GEO) orbit with a ~10° of magnetic latitude, and data used here range from 1999 to the beginning of 2008. The Synchronous Orbit Particle Analysis (SOPA) instrument [Belian et al., 1992] measures directional electron distributions from 50 keV to 1.5 MeV. With no magnetic instrument aboard, the local magnetic field direction is indirectly derived from the symmetries of the plasma distribution [Thomsen et al., 1996; also see Chen et al., 2005, Appendix A]. Extra care has been taken to remove
contaminations from background and other possible sources by fitting measurements with two-Maxwellian distributions [Cayton et al., 1989]. Flux data here have a time resolution of 10 min. Figure 2e presents one SOPA data example, showing pitch angle resolved fluxes for 1368 keV electrons on 1 day. Daily variations in PADs can be clearly seen with significant pancake PADs around local noon at ~0600 UT.

For all three data sets, the quiet Olson and Pfitzer magnetic field model [Olson and Pfitzer, 1977] together with the International Geomagnetic Reference Field (IGRF) are used for calculating the McIlwain’s L shell [McIlwain, 1966], MLT values, and mapping local PADs down to the magnetic equator. We choose to use McIlwain’s L instead of Roederer’s L* just for the reason of simplicity and convenience (see supporting information for more discussion). For developing REPAD, we select only local PADs measured within ± 15° magnetic latitude. This is a reasonable range picked after testing several values: Larger magnetic latitude ranges will leave the unmeasured pitch angle gap in the equatorial PAD too wide that we will not be able to capture some interesting features around the 90° pitch angle, and unreliable extrapolation is also unavoidable; smaller magnetic latitude ranges will discard many observations and make sample sizes too small to make the statistical sense. Inter calibration between instruments and data flagging have been performed in previous work [Friedel et al., 2005] and applied here. For model development, CEEPAD and SOPA fluxes are all first interpolated to MEA’s energy channels.

3. Pitch Angle Fitting Scheme and Model Formalism

One prerequisite for developing an empirical PAD model is an effective fitting method to quantify various PADs using limited parameters. Some previous studies have used the function form of \( \sin^n(\alpha) \), where \( \alpha \) is the pitch angle and \( n \) is the fitting parameter. This fitting function is good for presenting monotonically increasing and decreasing PADs between 0° and 90° pitch angles, as shown in previous studies [e.g., Vampola, 1998; Gannon et al., 2007], but obviously has difficulty in fitting some PADs, such as butterfly distributions. This means that one parameter is not enough for this work and what we need is a full set of orthogonal functions. If we use the full set of trigonometric functions, however, the problem is that we may end up with too many fitting coefficients that are difficult to handle and interpret.

Since a PAD is indeed the distribution over a sphere with the inclusion of azimuthal angles, using the spherical harmonics is an appealing solution, and we thus select Legendre polynomials as the fitting function set. Here we focus on explaining the fitting scheme, and details about these polynomials \( P_n(\cos(\alpha)) \) can be found in the supporting information.

\[
\ln|f(\alpha)| = \sum_{n=0}^{\infty} C_n P_n(\cos(\alpha)) = |C_0| \sum_{n=0}^{\infty} c_n P_n(\cos(\alpha))
\]  

(1)

As shown in equation (1), for a given pitch angle distribution \( f(\alpha) \), we choose to fit to the logarithm of fluxes \( j/|\alpha| \), which is why we use the logarithmic form later. The orthogonal relationship between polynomials enables us to calculate fitting coefficients \( C_n \) (c in upper case), and then we derive the normalized coefficients \( c_n \) (c in lower case),

\[
c_n = \frac{C_n}{|C_0|} = \frac{2(2n + 1)}{(2n + 1)!} \int_0^{\pi} \ln|j(\alpha)|*P_n(\cos(\alpha))*\sin(\alpha) d\alpha
\]

(2)

as defined in equation (2). Since \( |C_0| \) means the directionally average flux level for a distribution, coefficients \( c_n \) reflect the relative variations over the average level.

To help understanding the advantages of this fitting scheme, Figure 3 (top) plots several polynomials. First, trapped electrons request symmetric PADs over 90° pitch angle, which can be easily identified by coefficient values for the odd-th orders of \( P_n \), and thus, all fittings with large values in \( c_1, c_3, \) etc. are questionable and consequently excluded from model samples. Second, shapes of \( P_2 \) (red) and \( P_4 \) (blue) are alike the typical pancake and butterfly PADs, upside down though, suggesting the values of the first several even-th orders of \( P_n \) may dominate others. This has been demonstrated by fitting observations throughout the CRRES mission, as one example in Figure 3 (bottom), and showing that \( c_4 \) quickly decreases with increasing \( n \). Therefore, this justifies to keep \( c_0, c_2 \) and \( c_4 \) being the only three fitting coefficients for REPAD. In fact, our discussions next will only include \( c_0, c_2 \) and \( c_4 \) due to their much larger values. Finally, since the high even-th orders of \( P_n \) involves high-frequency fluctuations, e.g., the \( P_{20} \) curve (dashed black line) shown in Figure 3 (top), large values in the corresponding coefficients can be used for ruling out observations with poor statistics, which occur.
Often for high-energy electrons. Therefore, advantages of Legendre polynomials make them appropriate for fitting particle PADs, particularly in automated processing of large volume of data. In addition, this is also an effective and efficient fitting scheme. For example, through the whole CRRES mission, there are ~98% (72%) observations for electrons with the lowest energy 148 keV (highest energy, 1582 keV) that can be fitted to a fairly high degree with the only three fitting coefficients. This helps rule out the possibility of data-preselecting bias introduced by the fitting scheme. Figure 1 already shows several fitting results, and extra examples for good and poor fitting can be found in Figure S1 in the supporting information that are randomly selected from fitting MEA data. Data with poor fitting are excluded from REPAD. Values of coefficients $c_n$ can be directly read for categorizing PAD types, and such examples are shown in Figure 4 that plots PADs corresponding to pairs of $(c_2, c_4)$. Pancake PADs usually have negative $c_2$ and smaller values of $|c_4|$, such as those in Figure 4e; flattop PADs have both negative and comparable $c_2$ and $c_4$ values, such as those gray and green curves in Figure 4f; butterfly PADs have negative and large $c_2$ and smaller/comparable values of $|c_4|$, such as those in Figures 4g and 4h. In other words, by knowing the values of $(c_2, c_4)$, one may immediately know the type of PADs. This is also significant for automated processing, e.g., picking out PADs with some specific types from large volume of data sets and thus facilitates identifying statistical relation between some PAD shape and other physics parameters such as geomagnetic activity indices.

It must be mentioned that using Legendre polynomials for fitting PADs is not a new idea. This method first appeared in the work by Higbie and Moomey [1977] and was later applied for analyzing GEO electron data in case studies [e.g., Baker et al., 1978]. However, it is this work that for the first time thoroughly checks the method's effectiveness and further applies it to the development of a statistical model.

Now we explain the formalism of REPAD. For trapped radiation belt electrons, owing to Liouville's theorem, an off-equatorial PAD can always be magnetically converted to part of the equatorial PAD. Therefore, the spatial domain of REPAD is confined to the 2-D magnetic equator divided by many spatial bins with unequaled size. The center of each bin has the coordinates of (L, MLT), i.e., the Mcllwain L shell and magnetic local time, as plotted in gray contours in Figure 5a. In addition, eight energy points, ranging from 148 up to 1582 keV, are used as the dimension to reflect energy dependence. Meanwhile, to reflect the PAD's dependence on magnetic activities, the model is further sorted into three $AE$ categories. (Note that our selection of $AE$ is simply based on the consideration of electron's cyclotron resonance with waves. There are other parameter options such as the $Dst$ index, a representative of the changing global magnetic
configuration that affects electron PADs. Here we leave this topic for future studies.) As mentioned above, we only keep three fitting coefficients, $c_2$, $c_4$, and $c_6$ for PAD shapes. Finally, in each model bin, we keep the occurrence distribution for each $c_n$ as the example shown in Figures 5b and 5c. This distribution enables one not only to know the mean and median fitting values but also to estimate the error range. For example, in the

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Figure 4. Shapes of PADs corresponding to selected $(c_2, c_4)$ combinations. The central plot shows selected $(c_2, c_4)$ pairs in various colors and magnitudes. For pairs in each radial direction, the corresponding PAD curves are plotted as a function of $\cos(\alpha)$ in the same color in a small panel, distributed along the outermost circle along the same radial direction. All other fitting coefficients are assumed zero except for $c_0$ being one here. For example, pairs along the negative horizontal axis have PAD curves as shown in Figure 4e. (d–h) Panels marked with the bold red fonts are what typically observed in data, and (a–c) all others are much less commonly observed.

Figure 5. Model grids used in REPAD. (a) Contours of L shell grids (gray circles) and MLT grids (radial gray lines) in the magnetic equatorial plane, which are overplotted on the distribution of data sample numbers for 782 keV electrons within the AE range of [0, 100] nT. (b) Occurrence distribution of the normalized fitting coefficient $c_2$ in one bin (L = 7.0, MLT = 0) selected from Figure 5a. The vertical gray lines indicate $c_2$ values at five accumulative percentiles from the tenth to ninetieth with an increment of 20, and the red vertical line indicates the mean $c_2$. (c) Values of $c_2$ versus occurrence percentiles for the same model bin.
specific bin as shown in Figure 5c, the median $c_2$ has the value of $\sim 0.1$ at the 50th percentile, and the variation ranges between $\sim 0.0$ and 0.9 from the 10th to the 90th percentiles. This model requests the minimum sample number of 100 in each grid bin to keep the distribution statistically meaningful. A summary of grid points for REPAD is presented in Table 1.

Finally, since coefficients $c_n$ are treated as separate parameters in the model, the independence between each other as well as to the average flux level $C_0$ need to be confirmed. According to the statistics theory, two events $X$ and $Y$ are independent if and only if their joint probability, i.e., $\text{Prob}(X, Y)$, equals the product of their individual probabilities, i.e., $\text{Prob}(X) \times \text{Prob}(Y)$. Here we present examples for 782 keV electrons observed during the whole CRRES mission, as shown in Figures 6 and 7. Figures 6 and 7 (left column) show joint probabilities derived directly from observations, and Figures 6 and 7 (right column) show products of two individual probabilities, derived by collapsing 2-D distributions shown in left to each single dimension.

### Table 1. Summary of Grid Points in REPAD

<table>
<thead>
<tr>
<th></th>
<th>Values</th>
</tr>
</thead>
<tbody>
<tr>
<td>L shell grid points (total number 22)</td>
<td>3.0, 3.2, 3.4, 3.6, 3.8, 4.0, 4.2, 4.4, 4.6, 4.8, 5.0, 5.25, 5.5, 5.75, 6.0, 6.25, 6.5, 6.75, 7.0, 7.25, 8.0, 9.0</td>
</tr>
<tr>
<td>Magnetic local time grid points (in hour, total number 12)</td>
<td>00, 02, 04, 06, 08, 10, 12, 14, 16, 18, 20, 22</td>
</tr>
<tr>
<td>Energy grid points (in keV, total number 8)</td>
<td>148; 272; 509; 782; 1,090; 1,368; 1,472; 1,582</td>
</tr>
<tr>
<td>Three AE categories (in nT)</td>
<td>[0, 100); [100, 300); [300, 20,000]</td>
</tr>
<tr>
<td>Occurrence percentiles for $c_n$'s (total number 11)</td>
<td>0th, 10th, 20th, …, 90th, 100th</td>
</tr>
</tbody>
</table>

Figure 6. Check the independence between fitting coefficient $C_0$ and $c_n$. (a1) The joint probability distribution of $(C_0, c_2)$, i.e., $\text{Prob}(C_0, c_2)$, by fitting PADs for 782 keV electrons as observed during the whole CRRES mission. (a2) The product of two individual probabilities, i.e., $\text{Prob}(C_0) \times \text{Prob}(c_2)$, as calculated from Figure 6a1. (b1, b2) Distributions for $C_0$ and $c_4$ in the same format. (c1, c2) Distributions for $C_0$ and $c_6$. 

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Comparing two distributions in each row, one can easily see their similarities. Results for other energies have similar results (not shown here). Therefore, we claim that the usage of $c_n$ as independent model parameters is justified.

4. Model Results and Discussions

This section illustrates distributions from REPAD, though it is difficult to present many details due to the multiple model dimensions. Here we select and present some typical and representative examples, and more can be found in the supporting information. In addition, we also have some initial discussions on possible reasons leading to observed PADs, such as variations for different $AE$ categories, but pursuing definitive conclusions will need further detailed studies and thus are left to the future as a separate topic.

First, we present global distributions for the median values of fitting coefficients $c_n$, which determine PAD shapes and are the most widely used. Figure 8 shows examples for electrons with a moderate energy of 782 keV. One significant feature in these distributions is the difference between two distinct regions separated by L shell at ~6.5. Comparing panels for the two $AE$ categories, one can clearly see large variations of $c_n$ in the region outside of $L \sim 6.5$ while not much inside. (The inner most region with $L < 4$ also show variations that will be discussed later.) Similar two separated regions are also seen for low- and high-energy electrons (see Figure S2 in the supporting information). This should indicate that electrons in the two regions experience different level of adiabatic effects, which is at least partially related to $AE$, and/or other different physical processes, e.g., magnetopause may indeed reach $L \sim 6.5$ quite often but rarely inward.

For moderate-energy electrons, Figure 8 also shows that the relatively stable PADs in the inside region always stay close to isotropic since all $c_n$ values are very close to zero, while PADs in the outside region are very different and are strongly asymmetric between the dayside and nightside. The same can be read from curves

![Probability distribution](image)
During quiet times (i.e., the lowest AE category), PADs in dayside are pancake like, by using plots in Figure 4, and those at nightside are butterfly like; while for the highest AE category PADs will keep their typical forms but turn to be more isotropic around 90° pitch angle. This asymmetry between day and nightside can be partially explained by drift-shell splitting due to asymmetric magnetic field, while other processes (i.e., in situ wave-particle resonance) may also play a role here. PADs for low-energy electrons also turn more isotropic at nightside for high AE (Figure S2a), which is consistent with previous results that substorm injected electrons are almost isotropic \cite{Asnes et al., 2005}. High-energy electrons show similar asymmetry though the dayside pancake PADs turn to be more anisotropic, for which one explanation is due to the steep radial gradient in flux distributions during disturbed periods.

For each bin in the outside region, statistical distributions of $c_n$ exhibit large variations as shown in Figure 9. For example, $c_2$ and $c_4$ in the outside region at $L < 4$ show large variations. These variations are seen for high-energy electrons but not for the low-energy electrons (not shown here). This may reflect the resonance effectiveness of hiss waves inside the plasmapause that is much higher with MeV electrons than $\sim 100$ keV electrons \cite[e.g.,][]{Meredith et al., 2007}. Small variations for the region in between may reflect the dominance of some process with weak pitch angle dependence such as radial diffusion. Focusing on the 00 MLT, as shown in Figure 10, we see that coefficients $c_n$ in the outside region have the tendency to be closer to zero for higher AE values, which reflects
some process that works across all pitch angles similarly and thus make PADs more isotropic; while for L < 4, coefficients $c_n$ for moderate- and high-energy electrons turns more negative, which may reflect the hiss wave resonance being stronger for small pitch angles [Meredith et al., 2009]. No matter what is the physics behind, the larger variation in $c_n$ obviously indicates less representative the median/mean PAD distribution and vice versa. Also can be seen from Figure 9 is that the median values (in aqua color) are close to but not overlapping with the mean values (red), indicating occurrence distributions are not normal distributions.

Since REPAD uses three different data sets, it is necessary to check the difference between submodels that are based upon each individual data set. As shown in Figure 11, one can see that submodels agree with each other quite well in most regions, though differences do exist such as at L shells < ~4.5 for moderate-energy electrons (Figure 11, middle row). The differences may come from different instrument performance and/or different mission periods. Since Polar has the largest size of data, REPAD distributions are mostly dominated by Polar observations, except for the GEO region where LANL-97A also has significant numbers of observations.

5. Model Verification and Validation

We verify and validate the model using both in-sample and out-of-sample tests. The in-sample tests can verify that there are no significant algorithm and coding errors in the model development steps. For this purpose, we test the model by comparing individual CRRES, Polar, and LANL-97A data to model fitting results.
From observations shown in Figure 2, for example, we pick fluxes observed at local 90° pitch angles only and feed them to the model, and the fitting results are shown in Figure 12. Figures 12a, 12c, and 12e present fitted fluxes as a function of pitch angle and time, and Figures 12b, 12d, and 12f present the flux ratios between fitted results and observations. Clearly, the model always captures the characteristic shapes of PADs, and fitted fluxes agree with observations fairly well, except for pitch angles close to loss cones at some time. Extra gaps in fitted fluxes come from the L shell limitation of REPAD.

Figure 10. REPAD distributions of $c_n$ as a function of energy, L and AE. In each panel, the mean $c_n$ curves at 00 MLT are plotted as a function of L shell and colors indicate different AE categories. (top to bottom) Two rows correspond to $c_2$ and $c_4$, respectively. (left to right) Three columns correspond to three energies: 148, 782, and 1472 keV, respectively.

Figure 11. Cross-check fitting coefficients from three submodels. In each panel, $c_n$ is plotted as a function of L and data points with different colors are from different submodels: Orange for Polar, green for CRRES, blue for LANL-97A, and black curves for REPAD. (top to bottom) Three rows are for $c_2$ for 148 keV electrons in the lowest AE category, $c_4$ for 782 keV electrons in the medium AE category, and $c_6$ for 1582 keV electrons in the highest AE category, respectively. (left to right) Four columns are for four MLTs: 00, 06, 12, and 18 MLT.
For the out-of-sample test, we use the independence data from Van Allen Probes (formerly named RBSP). The two probes a and b have nearly identical, highly elliptical, and low-inclination orbits crossing radiation belt every ~9 h, with the apogee inside of GEO orbit. The Magnetic Electron Ion Spectrometer (MagEIS) instrument on board each measures electrons and ions in the middle energy ranges, and the pitch angle resolved fluxes we have here are from the M75 sensor covering the energy range of ~300–1000 keV [Blake et al., 2013]. Background contaminations on MagEIS from Bremsstrahlung and penetrating protons are not completely removed from data used here, and we assume they have low levels with pitch angle independence and thus have insignificant effects on the measured PADs particularly in the outer belt region. One example is shown for electrons with moderate energy (584 keV) on one magnetically active day, as shown in Figure 13. Again, here we use fluxes at the local 90° pitch angle only as input for REPAD to fit fluxes at other pitch angles. Flux values and ratios shown in Figures 13b and 13c indicate great model performance. Tests on other energies (Figure S5 for low energy and Figure S6 for high energy in the supporting information) for the same day also show similar results. Therefore, this out-of-sample test demonstrates the validity of REPAD.

To systematically evaluate the performance of REPAD, we further use 2 month observations from Van Allen Probes b (October–December 2012) and repeat above steps for each available energy channel. The
model's performance is evaluated in two methods: One is to get the average flux ratios as a function of energy and L shell, as shown in Figure 14a. It can be seen that for most model regions the ratio is close to 1 except for L ~3.5 and energies above 600 keV. This method show how REPAD works in the average sense. The other method is to calculate the performance efficiency (PE) for each model bin so as to evaluate the model's performance from the statistical sense. Here PE = 1 indicates model results perfectly fit

$$PE = 1 - \frac{\sum [\log(j_m) - \log(j_d)]^2}{\sum [\log(j_m) - \log(j_d)]}$$

where subscripts $m$ and $d$ for flux $j$ indicate fitted and observed values, respectively, and $\bar{j_d}$ is the mean of observations. Here PE = 1 indicates model results perfectly fit

Figure 13. REPAD Validation: Out-of-sample test. (a) Pitch angle resolved fluxes for 584 keV electrons observed by RBSP-b MagEIS instrument on 1 October 2012. Measurements with L < 3 are omitted here due to the L shell coverage of REPAD. (b) Fitted pitch angle resolved fluxes for the same energy, derived from REPAD using unidirectional fluxes at 90° local pitch angle from observations. (c) Ratios between fitted and observed fluxes as in Figures 13b and 13a. (d) Curves for L shell (black) and magnetic field ratio between local and equatorial values (gray). Equatorial pitch angle values (black) corresponding to five magnetic ratios are also given to the right. (e) MLT trace (black) for the satellite and AE index (gray).

Figure 14. REPAD performance evaluations. By comparing two months Van Allen Probes electron observations to fitted fluxes, we here plot (a) averaged flux ratios and (b) model prediction efficiency on logarithm of fluxes $\log(j)$ as a function of L and energy.
channels). For each energy channel, given the GPS L shell and MLT at one time point, REPAD provides geomagnetic activities. Extrapolation is needed for both L shell (beyond 9) and energy (the highest three energy categories, dark gray curves are for originally median $c_2$, downshifted by 0.5 from original values, and light gray curves are for $c_2$ upshifted by 0.5. (b1, b2) Occurrence distributions of flux ratios for 782 keV electrons observed by CRRES. The ratio is between the flux of equatorially mirroring electrons and locally directionally averaged flux, measured by CRRES but along simulated GPS orbits, at two L shells: ~GPS equatorial crossings (Figure 15b1) and ~GEO (Figure 15b2). In each panel, red vertical line indicates the mean ratio, while two blue lines for ratios at the 5th and 95th occurrence percentiles.

First, we discuss characteristic electron PADs at two L shells: the GPS equatorial crossing points at $L \sim 4.2$ and GEO at $L = 6.5$. As pointed out in section 4, at $L = 4.2$, fitting coefficients $c_4$ are highly stable, showing very small variations and being very close to zero as can be seen from Figure 15a1, while large variations in $c_6$ are seen around the GEO orbit (Figure 15a2). This implies that it is reasonable to assume electrons observed at $L = 4.2$ have fixed PADs. This is further tested by inspecting into CRRES electron data, which are pitch angle resolved and measured close to the equator. Flying a virtual GPS satellite through the CRRES data set, we are able to simulate its observed omnifluxes. Dividing the omnifluxes by $4\pi$, we have the directionally averaged fluxes observed by the virtual satellite. Comparing the directionally averaged flux to the model data, we can test the assumption of isotropic PADs. From Figure 15b1, the narrow distribution of flux ratios at $L = 4.2$ during the whole CRRES mission demonstrates that assuming isotropic PAD will bring an error factor between 1 and 1.5 for 782 keV electrons, while the error factor ranges beyond 0.1–10 for $L = 6.5$ (~GEO). Similar results are seen for low- and high-energy electrons (Figure S8 in the supporting information). Therefore, the assumption of isotropic PADs works quite well at GPS equatorial crossings, as we did in our previous electron phase space density studies [Chen et al., 2007a], but not so well at GEO orbit.

Bearing in mind that error bars grow at larger $L$ shells, we apply REPAD model to GPS observations throughout the whole outer belt region. Here one example is shown in Figure 16, which is on 1 day with moderate geomagnetic activities. Extrapolation is needed for both L shell (beyond 9) and energy (the highest three energy channels). For each energy channel, given the GPS L shell and MLT at one time point, REPAD provides fitting coefficients $c_n$ that determine the equatorial PAD shape. By finding out the $c_n$ that can best fit the directionally averaged flux from the observation, we are able to derive the full equatorial pitch angle distribution of fluxes from equations (1) and (2). Knowing the satellite’s geomagnetic latitude, we can derive the local pitch angle.

**Figure 15.** Variations of PADs at two typical L shells. (a1) REPAD fitting coefficients at GPS equatorial crossings with $L = 4.2$ as a function of energy, compared to (a2) those at ~GEO with $L = 6.5$. In each panel, black curves are for original median $c_2$ values for all local times and AE categories, dark gray curves are for $c_2$ downshifted by 0.5 from original values, and light gray curves are for $c_2$ upshifted by 0.5. (b1, b2) Occurrence distributions of flux ratios for 782 keV electrons observed by CRRES. The ratio is between the flux of equatorially mirroring electrons and locally directionally averaged flux, measured by CRRES but along simulated GPS orbits, at two L shells: ~GPS equatorial crossings (Figure 15b1) and ~GEO (Figure 15b2). In each panel, red vertical line indicates the mean ratio, while two blue lines for ratios at the 5th and 95th occurrence percentiles.

**6. Application to GPS Data**

As mentioned in section 1, one application of REPAD is for GPS electron data. Each GPS satellite has a circular orbit with a radius of 4.2 $R_E$, inclination of 55°, 12 h period, and crosses the outer electron belt every 6 h at different local times. Since GPS satellites carry only omnidirectional particle sensors, e.g., the Burst Detector Dosimeter IIR [Cayton et al., 1998] onboard ns41, adding pitch angle information to GPS electron observations has significant scientific meaning for radiation belt research.
distribution as shown in Figures 16c–16j. More examples on geomagnetically active and quiet days can be found in Figures S9 and S10 in the supporting information. A processed GPS data set, with fitted pitch angle distribution of fluxes and known error ranges, can be useful for data assimilation models for radiation belts.

7. Summary

We completed a statistical survey on the pitch angle distributions of energetic electron trapped in the Earth's outer radiation belt, and based upon those survey results we developed a new empirical model. This model—REPAD—is able to provide statistical pictures of electron equatorial pitch angle distributions as a function of energy, L shell, magnetic local time, and magnetic activity. To quantify and facilitate this statistical survey, we used Legendre polynomials for fitting long-term in situ directional fluxes observed near the magnetic equator from three missions: CRRES, Polar, and LANL-97A. Covering the whole outer belt region with energies up to 1.6 MeV, REPAD provides not only the mean and median pitch angle distributions but also error bars, which is a new and useful feature that has not existed before. Preliminary verification and validation results demonstrate the model construction, and tests against Van Allen Probes observations validate the model's reliable performance. Usage of REPAD is mainly to predict the full pitch angle distribution of fluxes along a given magnetic field line, or even on a given drift shell, based upon a single-point flux measurement on
that line with limited directional resolution (or not at all). In particular, the very small variations in pitch angle distributions measured with $L$ within $\sim 4-5$ justify the assumption of fixed pitch angle distributions at GPS equatorial crossings ($L \sim 4.2$).

References


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