

He Fei (Orcid ID: 0000-0003-0542-2686)
Zhang Xiaoxin (Orcid ID: 0000-0002-7759-7402)
Fok Mei-Ching, H. (Orcid ID: 0000-0001-9500-866X)
Liemohn Michael, W. (Orcid ID: 0000-0002-7039-2631)
Gallagher Dennis (Orcid ID: 0000-0003-3924-3571)
Nakano Shin'ya (Orcid ID: 0000-0003-0772-4610)

A new solar wind driven global dynamic plasmopause model: 2. Model and Validation

Fei He, Key Laboratory of Earth and Planetary Physics, Institute of Geology and Geophysics, Chinese Academy of Sciences, Beijing, China (Email: hefei@mail.iggcas.ac.cn)

Xiao-Xin Zhang, Key Laboratory of Space Weather, National Center for Space Weather, China Meteorological Administration, Beijing, China (**Corresponding author, Email: xxzhang@cma.gov.cn**)

Rui-Lin Lin, National Space Science Center, Chinese Academy of Sciences, Beijing, China (Email: linrl@nssc.ac.cn)

Mei-Ching Fok, NASA Goddard Space Flight Center, Code 673, Greenbelt, MD 20771, U.S.A. (Email: mei-ching.h.fok@nasa.gov)

Roxanne M. Katus, Department of Mathematics, Eastern Michigan University, Ypsilanti, Michigan, USA (Email: rkatus@emich.edu)

Mike W. Liemohn, Department of Climate and Space Sciences and Engineering, University of Michigan, Ann Arbor, Michigan, USA (Email: liemohn@umich.edu)

Dennis L. Gallagher, NASA Marshall Space Flight Center, Huntsville, Alabama, U.S.A. (Email: dennis.gallagher@nasa.gov)

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Shinya Nakano, The Institute of Statistical Mathematics, Tachikawa, Tokyo, Japan (Email: shiny@ism.ac.jp)

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Abstract.

A new solar wind driven global dynamic plasmopause (NSW-GDP) model has been constructed based on the largest currently-available database containing 49119 plasmopause crossing locations and 3957 plasmopause profiles (corresponding to 48899 plasmopause locations), from 18 satellites during 1977 – 2015 covering four solar cycles. This model is compiled by the Levenberg-Marquardt method for nonlinear multiparameter fitting and parameterized by V_{sw} , B_z , $SYM-H$, and AE . Continuous and smooth MLT-dependence controlled mainly by the solar wind driven convection electric field E_{sw} is also embedded in this model. Compared with previous empirical models based on our database, this new model improves the forecasting accuracy and capability for the global plasmopause. The diurnal, seasonal and solar cycle variations of the plasmopause can be captured by the new model. The NSW-GDP model can potentially be used to forecast the global plasmopause shape with upstream solar wind and IMF parameters and corresponding predicted values of $SYM-H$ and AE , and can also be used as input parameters for other inner magnetospheric coupling models, such as dynamic radiation belt and ring current models and even MHD models.

Key Points:

1. A new solar wind driven global dynamic plasmopause model based on multi-satellite observations is constructed
2. This model is parameterized by V_{sw} , IMF B_z , $SYM-H$, and AE , and has continuous and smooth MLT-dependence

3. This model is potentially applicable to inner magnetospheric research studies and space weather forecasts

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1. Introduction

The plasmopause is the outer boundary of the plasmasphere, where the plasma density drops dramatically by at least half an order of magnitude in a short distance of $\sim 0.5 R_E$ (Earth's radii, $1 R_E = 6378.0$ km) [Carpenter, 1963; Gringauz, 1963]. The plasmopause configuration is one of the key parameters in the coupling interaction of the plasmasphere, the ring current, and the radiation belts in the magnetosphere and is also an important indicator for geomagnetic activity in near-Earth space [Carpenter and Anderson, 1992; Fok et al., 1995; Khazanov and Liemohn, 1995; Goldstein et al., 2003a; Spasojević et al., 2003].

The plasmopause locations can be characterized by physics-based models or statistics-based empirical models. For the physics-based models, some are based on a fluid approach, such as the dynamic global core plasma model (DGCPM) [Ober et al., 1997; Liemohn et al., 2004; He et al., 2013] in which the plasmopause is identified from the radial steep gradient in the density profiles, and the others are based on a kinetic approach, such as the dynamic kinetic model of the plasmasphere by Pierrard and Stegen [2008] in which the plasmopause is derived from zero parallel force (ZPF) surface [Lemaire, 1989]. For a detailed review of the physics-based models of the plasmasphere, please refer to Pierrard et al. [2009]. After the review of Pierrard et al. [2009], a physics-based reconstruction of the density in the plasmasphere was presented by Verigin et al. [2012] and Kotova et al. [2015], where the plasmopause was described as the last closed stream line but with rather flexible shape.

Many statistics-based empirical models of the plasmopause have been developed in past studies.

The functional representation of the geocentric radius of the plasmapause (L_{PP} , in R_E) was first proposed by *Carpenter and Anderson* [1992] (hereafter, this model is referred to as the CAA-1992 model). The CAA-1992 model is a Kp -based empirical model with $L_{PP} = 5.6 - 0.46 Kp_{\max}$, where Kp_{\max} is the maximum Kp value in the preceding 24 hours, and where the model was developed for the 0 h – 15 h magnetic local time (MLT) sector based on plasmapause crossing events taken from International Sun-Earth Explorer (ISEE-1) data in 1977, 1982 and 1983. Following the CAA-1992 model, *Gallagher et al.* [2000] constructed the empirical global core plasma model (GCPM) to characterize the core plasma density and composition in the inner magnetosphere. This is the first model to provide smooth, continuous total density globally.

A new Kp -dependent model of $L_{PP} = (5.39 \pm 0.072) - (0.382 \pm 0.019) Kp_{\max}$, where Kp_{\max} is the maximum Kp value in the preceding 12 hours, was proposed by *Moldwin et al.* [2002] (hereafter referred to as the MOL-2002 model) based on Combined Release and Radiation Effects Satellite (CRRES) measurements in 1990-1991. Using the same database of *Moldwin et al.* [2002], *O'Brien and Moldwin* [2003] further developed an MLT-dependent L_{PP} model with the following formula (hereafter referred to as the OBM-2003 model):

$$L_{PP} = a_1 [1 + a_{\text{mlt}} \cos(\Phi - a_\Phi)] Q + b_1 [1 + b_{\text{mlt}} \cos(\Phi - a_\Phi)] \quad (1)$$

where $Q = \max_{-36,-2} Kp$, $Q = \log_{10} \max_{-36,0} AE$, or $Q = \log_{10} |\min_{-24,0} Dst|$, $\Phi = 2\pi(\text{MLT}/24)$, a_1 , a_{mlt} , a_Φ , b_1 , b_{mlt} , and b_Φ are fitted parameters. The notation $\max_{t_1,t_2} X$ or $\min_{t_1,t_2} X$ indicates the maximum or minimum of X taken from t_1 to t_2 hours before the plasmapause crossing. Recently, *Bandić et al.* [2016] revisited the same CRRES data and constructed a new model (BAN-2016) similar to

OBM-2003 but with different coefficients.

The plasmopause locations extracted from the Imager for Magnetosphere-to-Auroral Global Exploration (IMAGE) Extreme Ultraviolet Imager (EUV) images were used by *Larsen et al.* [2007] to construct the first solar wind driven plasmopause model (hereafter referred to as the LAR-2007 model) without considering MLT-dependence. The LAR-2007 model was a function of interplanetary magnetic field (IMF) B_Z component with a shift time of 155 min, and $\phi = V_{SW} B \sin^2(\theta_c/2)$ with shift time of 275 min for solar wind speed V_{SW} , IMF magnitude B , and IMF clock angle θ_c :

$$L_{PP} = 0.050 B_{Z,155} - 1.110 \times 10^{-4} \phi_{275} + 4.23 \quad (2)$$

A new fit function of L_{PP} to V_{SW} , B_Z and AE was recently proposed by *Cho et al.* [2015] (hereafter referred to as the CHO-2015 model) based on the plasmopause crossings from the Time History of Events and Macroscale Interactions during Substorms (THEMIS) during the ascending phase of Solar Cycle 24, achieving better prediction performance compared with previous models. *Liu et al.* [2015] used the THEMIS data to further construct an MLT-dependent L_{PP} model with input parameters of $SYM-H$, AL , AU , AE and Kp . Plasmopause crossings extracted from the Waves of High frequency and Sounder for Probing of Electron density by Relaxation (WHISPER) on Cluster were used by *Verbanac et al.* [2015] to construct the new L_{PP} formula as a function of B_Z , $V_{SW}B_Z$, $d\Phi_{mp}/dt$, Dst , Ap , and AE in three MLT sectors (1 h – 7 h, 7 h – 16 h, and 16 h – 1 h) (hereafter referred to as the VER-2015 model).

For the convenience of comparisons, all the above empirical models are summarized in Table 1. It is

shown that each of these models just used plasmaspheric observations from a single satellite, and the databases for these models were not large enough to fully and adequately cover the temporal-spatial-geomagnetic activity dimensions. Some of the models with MLT-dependence have just used geomagnetic indices as input parameters (e.g., LIU-2015), while other models using solar wind parameters have no MLT-dependence (e.g., LAR-2007 and CHO-2015). Since the plasmopause locations are highly MLT-dependent and also highly dependent on solar wind and geomagnetic activity, construction of a solar wind driven global dynamic plasmopause model with MLT-dependence is necessary.

In Paper 1 [Zhang *et al.*, this issue], the largest currently-available database containing 49119 plasmopause crossing locations and 3957 plasmopause profiles (corresponding to 48899 plasmopause locations in 1 h MLT intervals) was established for four solar cycles (21 – 24). Statistical characteristics of this database were discussed in detail in Paper 1, especially the diurnal, seasonal and solar cycle variations of the plasmopause. This database, with full MLT and solar activity coverage, provides a unique opportunity to construct the new solar wind driven global dynamic plasmopause (NSW-GDP) model in this investigation.

The paper will be arranged as follows. The correlations between the plasmopause locations and the solar wind and geomagnetic parameters will be investigated in section 2 to help us select the parameters important for controlling the size and the shape of the plasmopause. In section 3, we will get the relationships between the selected parameters and the plasmopause locations by the Levenberg-Marquardt method for nonlinear multiparameter fitting. In section 4, the NSW-GDP

model will be compared with previous models and relevant discussions will be given. Finally, a summary and conclusion will be presented in section 5.

2. Parameters Selection

As a core region of the inner magnetosphere, the dynamics of the plasmasphere is mainly controlled by two sources of drivers. The first is external convective driving from the solar wind and IMF [Goldstein *et al.*, 2003a, 2003b, 2005a, 2005b; Sandel *et al.*, 2003; Darrouzet *et al.*, 2009; Katus *et al.*, 2015; and references therein], which modifies large scale convection in the inner magnetosphere that drives the dynamic distribution of plasmaspheric plasma through $\mathbf{E} \times \mathbf{B}$ drifts. The second is internal driving due to the dynamics of magnetospheric energetic particles and the ionosphere [Goldstein *et al.*, 2003c, 2007; Liemohn *et al.*, 2004, 2006; He *et al.*, 2016; Zhang *et al.*, 2017], especially by auroral substorms that produce strong ion and electron precipitation in the ionosphere [Akasofu, 1964; McPherron *et al.*, 1973] and dipolarization in the magnetosphere [Runov *et al.*, 2009; Ge *et al.*, 2012]. In the following sections, we will investigate the correlations of the plasmopause location with geomagnetic indices and solar wind parameters to optimize the parameters that drive the NSW-GDP model.

2.1. Correlation with Geomagnetic Indices

The 3-hour averaged Kp index, the 1-hour averaged Dst index, the 5-minute averaged $SYM-H$ index, and the 5-minute averaged AE index are used in this section to investigate correlations of the plasmopause location to geomagnetic activity. Like previous studies [e.g., O'Brien and Moldwin, 2003; Cho *et al.*, 2015; Liu *et al.*, 2015], a time window of two days is set to find the best delay

time of the geomagnetic indices to plasmopause variations. We correlate the plasmopause locations with $\max_{t_1, t_2} Kp$, $\log_{10}(|\min_{t_1, t_2} Dst|)$, $\log_{10}(|\min_{t_1, t_2} SYM-H|)$, and $\log_{10}(\max_{t_1, t_2} AE)$, where t_1 and t_2 both varies from -48 h to 0 h with 0 h corresponding to the plasmopause crossing time, always keeping t_1 earlier than t_2 . The Spearman rank-order correlation coefficient (ROCC) [Press et al., 1992] of the three quantities to the plasmopause locations at different delay times are calculated and the times t_1 and t_2 corresponding to the strongest correlation are shown in Figure 1. It is indicated that the correlation is always negative for Kp , Dst , $SYM-H$, and AE . The correlation is always strong in the 0 h – 12 h and 21 h – 24 h MLT sectors and weak in the 12 h – 21 h MLT sector. The MLT-averaged values of t_1 are -26 h, -18 h, -18 h, and -18 h for Kp , Dst , $SYM-H$, and AE , respectively, and the corresponding t_2 values are -3 h, -1 h, -1 h, and -2 h. The Kp , Dst and $SYM-H$ indices are all good indicators of geomagnetic storms but with temporal resolutions of 3 hours, 1 hour, and 1 minute, respectively. In consideration of constructing a dynamic plasmopause model, it's better to use high resolution indices, though the ROCC's for these three indices are all high in Figures 1a-1c. Therefore, only the 5-minute averaged $SYM-H$ and the 5-minute averaged AE are used in our model given that $SYM-H$ is a good proxy of the geomagnetic storm and AE is a good indicator of the geomagnetic substorm. Inclusion of $SYM-H$ and AE into the model can improve the adaptability of the model since sometimes a geomagnetic disturbance can cause dynamic variations of the plasmopause even when the solar wind conditions are quiet and stable [Goldstein et al., 2007; Zhang et al., 2017].

2.2. Correlation with Solar Wind

For the solar wind and IMF parameters, since the OMNI data have been time shifted to the nose of the Earth's bow shock and the delay time from the bow shock to the inner magnetosphere has been considered in matching solar wind parameters to the plasmopause crossings, the time window method in the section above will not be used here. The solar wind parameters in our database will be directly correlated with the plasmopause locations. The ROCC's between L_{PP} and IMF B_X , B_Y , B_Z , $B_T = \sqrt{B_Y^2 + B_Z^2}$ (all in the GSM coordinate system in nT), IMF clock angle $\theta_c = \text{atan}(B_Y, B_Z)$, V_{SW} (km/s), N_{SW} (cm^{-3}), solar wind driven convection electric field $E_{SW} = 10^{-3} B_Z V_{SW}$ (mV/m), solar wind dynamic pressure $P_{dyn} = 2 \times 10^{-6} N_{SW} V_{SW}^2$ (nPa), and Akasofu's solar wind-magnetosphere coupling function $\varepsilon = V_{SW} B_T \sin^4(\theta_c/2)$ [Perrault and Akasofu, 1978] at different MLT sectors (in 1-h intervals) are calculated and plotted in Figure 2.

Characteristics in Figure 2 can be summarized as follows:

1. V_{SW} has the strongest correlation with L_{PP} compared with other parameters, with the ROCC being ~ -0.6 at all MLT sectors indicating a very strong negative correlation.
2. N_{SW} is positively correlated with L_{PP} though not so strong, with ROCCs around 0.2.
3. IMF B_Z and θ_c have the same correlations with L_{PP} . Although their ROCCs are symmetric about the line of ROCC=0, their physical means are the same since an increase of θ_c from 0° to 180° corresponds to a decrease of B_Z from positive to negative. Such correlations indicate that a southward turning of the IMF can cause the earthward shrinkage of L_{PP} .
4. Due to the strong correlation of V_{SW} with L_{PP} , the ε parameter, P_{dyn} , and E_{SW} , which are all related with V_{SW} according to their definitions, are all weakly correlated with L_{PP} with the

correlations becoming even weaker in the pre-dusk region around 16 h MLT. This is because correlation of B_Z with L_{PP} is the weakest and the positive correlation of N_{SW} to L_{PP} is the strongest in that region.

5. IMF B_X and B_Y are almost uncorrelated with L_{PP} at all MLT sectors.

Based on the above analysis, the two basic parameters, V_{SW} and IMF B_Z , which are the most important two parameters for solar wind-magnetosphere coupling and inner magnetospheric dynamics, are selected to be the input parameters of the NSW-GDP model. Therefore, in the following section, we will construct the NSW-GDP model driven by V_{SW} , B_Z , AE , and $SYM-H$.

3. Construction of the NSW-GDP Model

The basic framework of the model will be built on V_{SW} and B_Z in the first step. In the second step, the MLT-dependence of the plasmopause will be built on E_{SW} , and then AE and $SYM-H$ will be included in the model in the final step. Note that only 80% of the database is randomly chosen for modeling, the other 20% is used for comparison, and all calculations throughout the rest of this paper are performed in the magnetic equatorial plane of solar magnetic (SM) coordinates.

3.1. Basic Framework

To get the basic framework of the NSW-GDP model, it is better to eliminate the effects of MLT-dependence of the plasmopause. According to the statistics in Paper 1, the plasmopause shape in the 0 h to 6 h MLT sector is almost circular under many different solar wind and geomagnetic activity levels. Thus, we resample all the L_{PP} values from the 0 h to 6 h MLT sector to different V_{SW} bins with intervals of 50 km/s and different B_Z bins with intervals of 1 nT, respectively, to get the

dots in Figure 3 with standard deviations represented by the vertical error bars. It should be noted that plasmopause crossings with V_{SW} greater than 800 km/s and the magnitude of IMF B_Z greater than 15.0 nT are excluded in Figure 3 since the corresponding sample numbers are too small to ensure a reasonable statistical significance level for the t -test [Press *et al.*, 1992]. Best fitting of the dots in Figure 3a shows that the plasmopause decreases exponentially with V_{SW} as represented by the function in equation (3). As the magnitude of the IMF B_Z increases, the plasmopause shrinks towards the Earth, with the reduction more dramatic for southward IMF than for northward IMF. Such variations of L_{PP} with B_Z can be well depicted by an inverse polynomial function as shown by the black curve in Figure 3b. Therefore, the two fitted functions in Figure 3 are as follows with the fitted coefficients for $L_{PP}(V_{SW})$ and $L_{PP}(B_Z)$ listed in Table 2.

$$L_{PP}(V_{SW}) = f_0 + f_1 \exp(-V_{SW} / f_2) \quad (3)$$

$$L_{PP}(B_Z) = g_0 + \frac{g_1}{1 + g_2[(B_Z - g_3) / g_4]^2 + g_5[(B_Z - g_3) / g_4]^4 + g_6[(B_Z - g_3) / g_4]^6} \quad (4)$$

The two functions in Figure 3 are independent. It is expected that, for different IMF conditions, the coefficients for $L_{PP}(V_{SW})$ should change, and the same for $L_{PP}(B_Z)$ under different solar wind conditions. Therefore, establishment of a two-dimensional L_{PP} function of both V_{SW} and B_Z should be the most useful. To achieve this, we first present the B_Z variations of L_{PP} for different V_{SW} in Figure 4. Curves in Figure 4 are all best-fittings of equation (4) to averaged values in each V_{SW} interval. Note that the typical standard deviation for the symbols in Figure 4 is $\sim 0.5 R_E$. The t -test indicates that the maximum values at IMF $B_Z = +3.0$ nT in Figure 4 are significantly different with

each other with significance level of 99%, except for the comparison between the star and the diamond, for which a significance level of 95.5% is shown. For the symbols at IMF $B_Z = -15.0$ nT or $+15.0$ nT, the significance levels are all above 95%. This reveals that the symbols and curves in Figure 4 are truly different from each other.

It is demonstrated that the functional form of equation (4) can well depict the L_{PP} - B_Z relationship under different V_{SW} conditions. So, we have rewritten equation (4) in the following form:

$$L_{PP}(V_{SW}, B_Z) = g_0(V_{SW}) + \frac{g_1(V_{SW})}{1 + g_2(V_{SW})u^2 + g_5(V_{SW})u^4 + g_6(V_{SW})u^6}, \quad (5)$$

$$u = [B_Z - g_3(V_{SW})] / g_4(V_{SW})$$

where the seven V_{SW} dependent g functions can control the configurations of the fitted curves and the fitting method is as follows:

1. Divide the L_{PP} - B_Z data into different subsets according to the V_{SW} bins with an interval of 50 km/s;
2. Resample the L_{PP} data into B_Z bins with an interval of 1.0 nT;
3. Fit the data points to equation (4) to get $g_0, g_1, g_2, g_3, g_4, g_5,$ and g_6 for different V_{SW} , as shown in Figure 5.
4. Fit $g_0, g_1, g_2, g_3, g_4, g_5,$ and g_6 to V_{SW} to get the forms of the seven functions, respectively.
5. Adjust the coefficients of the seven functions iteratively to get an optimized function of equation (5).

After best-fitting with the Levenberg-Marquardt method [Press et al., 1992], the forms of the seven g functions are:

$$\begin{cases} g_0(V_{\text{SW}}) = g_{00} + g_{01} \exp(-V_{\text{SW}} / g_{02}), \\ g_1(V_{\text{SW}}) = g_{10} + g_{11} \exp(-V_{\text{SW}} / g_{12}), \\ g_2(V_{\text{SW}}) = g_{20} + g_{21} \exp(-V_{\text{SW}} / g_{22}), \\ g_3(V_{\text{SW}}) = g_{30} + g_{31} V_{\text{SW}}, \\ g_4(V_{\text{SW}}) = g_{40} + g_{41} \exp(-V_{\text{SW}} / g_{42}), \\ g_5(V_{\text{SW}}) = g_{50} + g_{51} \exp(-V_{\text{SW}} / g_{52}), \\ g_6(V_{\text{SW}}) = g_{60} \exp(-V_{\text{SW}} / g_{61}) \end{cases} \quad (6)$$

where the coefficients for these functions will be finally determined in section 3.4 with the bootstrap resampling method.

3.2. MLT-Dependence

In the previous section, we have established the functional representation of the L_{PP} by V_{SW} and IMF B_Z in the 0 h to 6 h MLT sector. The MLT variations of the plasmopause will be investigated to establish the MLT-dependence of the model. Since the MLT-dependence of the plasmopause is mainly controlled by the large-scale convection in the inner magnetosphere, the solar wind driven convection electric field E_{SW} , therefore, may be suitable to investigate the MLT-dependence of the plasmopause. Note that this statistical analysis cannot represent the often highly structured plasmopause profile that appears to result from meso-scale processes in the inner magnetosphere. While these structures can have profound radial and MLT variations, only their statistical properties are included.

Figure 6 shows the MLT-dependence of the L_{PP} under different conditions of E_{SW} . Firstly, the plasmopause locations are classified into six groups according to the values of E_{SW} as shown in the upper left corner of Figure 6. Secondly, the plasmopause locations in each group are binned into

each MLT sector with 1 h intervals. Finally, the averaged L_{PP} are obtained and plotted as colored dots in Figure 6. Each set of dots in Figure 6 is best fitted with the following function:

$$\begin{aligned} f(\text{MLT}) &= \frac{h_0 + h_1 \sin^2 \chi}{1 + h_3 \sin^2 \chi}, \\ \chi &= \frac{\pi(\text{MLT} - h_2)}{24} \end{aligned} \quad (7)$$

where h_0 , h_1 , h_2 and h_3 are functions of E_{SW} . After best fitting with the Levenberg-Marquardt method in each E_{SW} group, the forms of the four h functions are:

$$\begin{cases} h_0(E_{SW}) = h_{00} + h_{01}E_{SW} + h_{02}E_{SW}^2, \\ h_1(E_{SW}) = h_{10} + h_{11}E_{SW} + h_{12}E_{SW}^2, \\ h_2(E_{SW}) = h_{20} + h_{21}E_{SW}, \\ h_3(E_{SW}) = h_{30} + h_{31}E_{SW} + h_{32}E_{SW}^2, \end{cases} \quad (8)$$

where the coefficients for these functions will be finally determined in section 3.4.

3.3. Including Storm and Substorm Effects

As expressed in section 2, apart from the solar wind and IMF, the dynamics in the inner magnetosphere can also affect the shape and size of the plasmapause. The most significant dynamics in the inner magnetosphere are geomagnetic storms and substorms that can modify the convection patterns in the plasmaspheric region, thus change the shape and size of the plasmapause. As common proxies for geomagnetic activity, $SYM-H$ and AE are used to denote the storm and substorm activities, respectively. Therefore, we further attempt to add functions of $SYM-H$ and AE to the NSW-GDP model.

According to the statistics in section 2.1, the slopes of the linear fitting functions of L_{PP} to \log_{10}

$|SYM-H|$ and $\text{alog}_{10} AE$ at different MLT sectors are shown in Figure 7a and 7b, respectively. It is shown that both $SYM-H$ and AE have different effects on the plasmopause at different MLT sectors. For $SYM-H$, the slope decreases from 0 h to 10 h MLT and then increases until 16 h MLT when the slope experiences a decrease again. For AE , there is only one peak at 16 h MLT, around which the slopes are both decreasing.

According to the variations of the slopes in Figure 7, piecewise linear functions are adopted to characterize the relationship between L_{PP} and $SYM-H$ and that between L_{PP} and AE . Based on Figure 1, the MLT-averaged time window is $t_1 = -18$ h and $t_2 = -1$ h for both $SYM-H$ and AE . The function is:

$$\Delta L_{PP}(SYM-H, AE, MLT) = k_0 \text{alog}_{10} |\min_{-18,-1} SYM-H| + k_1 \text{alog}_{10} \max_{-18,-1} AE + k_2 \quad (9)$$

where the parameters k_0 , k_1 and k_2 are piecewise linear functions of MLT in the following forms:

$$k_0 = \begin{cases} k_{00} + k_{01}MLT, & 0 \leq MLT < 10, \\ k_{02} + k_{03}MLT, & 10 \leq MLT < 16, \\ k_{04} + k_{05}MLT, & MLT \geq 16 \end{cases} \quad (10)$$

$$k_1 = \begin{cases} k_{10} + k_{11}MLT, & 0 \leq MLT < 16, \\ k_{12} + k_{13}MLT, & MLT \geq 16 \end{cases} \quad (11)$$

$$k_2 = \begin{cases} k_{20} + k_{21}MLT, & 0 \leq MLT < 15, \\ k_{22} + k_{23}MLT, & MLT \geq 15 \end{cases} \quad (12)$$

The coefficients in equations (10) – (12) are to be determined in the final fitting of the model functions to the plasmopause location database.

Additionally, the contributions of storm or substorm activity to plasmaspheric evolution are different under different conditions of solar wind driven large scale convection. The plasmopause

locations are classified into four subsets based on the absolute values of E_{SW} , as listed in Table 3. The values of $dL_{PP-Substorm}$ and $dL_{PP-Storm}$ in Table 3 are obtained in the following manner. In each subset of E_{SW} , $dL_{PP-Substorm}$ equals the averaged plasmopause location during non-substorm periods with $AE < 200$ nT subtracted by the averaged plasmopause location during substorm periods with $AE > 500$ nT. For $dL_{PP-Storm}$, the subsets for non-storm and storm periods are $SYM-H > -20$ nT and $SYM-H < -50$ nT, respectively. It is demonstrated that as $|E_{SW}|$ increases, the contributions of both substorms and storms to the shrinkage of the plasmopause weaken almost linearly. Therefore, a switch function of $k_3(|E_{SW}|)$ has been added to the first two terms of equation (9), which is now rewritten as:

$$\Delta L_{PP}(SYM-H, AE, MLT) = \left\{ k_0 \log_{10} |\min_{-18,-1} SYM-H| + k_1 \log_{10} \max_{-18,-1} AE \right\} \times \left\{ k_{30} + k_{31} \tanh[(|E_{SW}| - k_{32}) / k_{33}] \right\} + k_2 \quad (13)$$

where the coefficients for the switch function will be determined in the following optimization procedure.

3.4. Determination of the Model Coefficients

Finally, the complete form of the NSW-GDP model has been constructed as follows:

$$L_{PP}(V_{SW}, B_Z, SYM-H, AE, MLT) = \left\{ g_0(V_{SW}) + \frac{g_1(V_{SW})}{1 + g_2(V_{SW})u^2 + g_5(V_{SW})u^4 + g_6(V_{SW})u^6} \right\} \times \frac{h_0(E_{SW}) + h_1(E_{SW}) \sin^2 \chi}{1 + h_3(E_{SW}) \sin^2 \chi} + \left\{ k_0(MLT) \log_{10} |\min_{-18,-1} SYM-H| + k_1(MLT) \log_{10} \max_{-18,-1} AE \right\} \times k_3(|E_{SW}|) + k_2(MLT), \quad (14)$$

$$u = [B_Z - g_3(V_{SW})] / g_4(V_{SW}),$$

$$\chi = \frac{\pi[MLT - h_2(V_{SW})]}{24},$$

where detailed definitions of the functions have been presented in previous sections. A bootstrap

resampling method [Efron and Tibshirani, 1993; Huber and Ronchetti, 2009], which has been used successfully before in space physics research to quantify the error on statistical results [e.g., Reiff, 1990; Liemohn et al., 2012; Katus et al., 2013], is conducted to determine the coefficients and test the robustness of the model. The bootstrap method is to randomly choose 80% of the database for the creation of the model, and then compare against the remaining 20%. For this investigation 1000 bootstrap resamplings are conducted to investigate the mean and spread of the coefficients in equation (14).

By fitting equation (14) to each resampling of the database in each MLT sector to maximize the ROCC and minimize the root-mean-square errors (RMSE) of the fitted plasmopause locations to the observed plasmopause locations, the coefficients for equations (14) are determined. Repeating the fitting procedure for 1000 times, the mean value and standard deviation (σ) of each coefficient in equation (14) are calculated and listed in Table 4. The standard deviations are generally within 5% of the mean values, indicating that the model fittings are robust. It is noted that, in equation (14), the upper limit values for V_{SW} , $|B_Z|$, and $|E_{SW}|$ are 800 km/s, 15.0 nT, and 10 mV/m, respectively, as analyzed in section 3.1.

4. Discussions on the NSW-GDP Model

4.1. General Performance

Comparisons between the observed plasmopause locations (L_{OBS}) and the model results (L_{MOD}) at different MLT sectors are shown in Figure 8 and the ROCC and RMSE are plotted in Figure 9. The error bars in Figure 9 indicate the standard deviations of the ROCC and RMSE calculated through

the bootstrap resampling method. Figure 8 is obtained in the following manner. In a specific MLT sector with N plasmopause locations, the L_{OBS} and corresponding L_{MOD} are resampled into two-dimensional grids with size of $0.2 R_E \times 0.2 R_E$, and the number of points in each grid is normalized by the total number of plasmopause locations N to get the normalized distributions in Figure 8. Generally, $\sim 70\%$ of the points are located in the range of $L_{\text{MOD}} \pm 0.5 R_E$, except for the 16 – 19 MLT sectors, the percentage is about $50\% \sim 60\%$. The ROCC of the NSW-GDP model is between 0.7 and 0.8 in 21 h to 12 h MLT sectors. In the afternoon to dusk region, a minimum value of ROCC = 0.56 is reached at 17 h MLT. The variation of RMSE with MLT is inversely changing with that of ROCC in Figure 9, indicating that a higher ROCC corresponds to a lower RMSE and thus a more successful modeling. The error bars in Figure 9 show that the model compilation method is robust regardless of the 80% of the database chosen to create the coefficient set.

4.2. Comparison with Previous Models

In this section, the NSW-GDP model will be compared with those models listed in Table 1. Based on the remaining 20% of the database, the RMSE of the various models listed in Table 1 are calculated and plotted in Figure 10. The most significant feature in Figure 10 is that the NSW-GDP model has the best performance in all MLT sectors compared with all other models. The maximum RMSE of NSW-GDP is $0.91 R_E$ at 17 h MLT sector, with a minimum of $0.57 R_E$ at midnight. The maximum RMSE of other models are all larger than $1.0 R_E$ with some models achieving $1.8 R_E$. Generally, the performances of all the models become poor in the afternoon to dusk MLT sectors. This might be caused by the fact that the plasmasphere is the most dynamic in the afternoon to dusk

MLT sectors where different structures of plumes may be generated [Moldwin *et al.*, 2004; Goldstein and Sandel, 2005; Darrouzet *et al.*, 2006, 2008; He *et al.*, 2011, 2013]. Besides, all the old models are based on plasmaspheric observations from single satellites in limited periods, and such models may be difficult to fit such a large database with a time span of nearly 40 years.

4.3. Modeled Plasmapause Variations

To analyze the variations of the plasmapause locations simulated by the NSW-GDP, e.g., the diurnal, seasonal, and solar cycle variations presented in Paper 1, the global plasmapause locations are calculated from 1980 to 2015 (a 36-year long interval) based on the hourly OMNI solar wind and IMF parameters and the geomagnetic indices. Then, the plasmapause locations are binned into an MLT-UT coordinate frame with 1 h intervals as shown in Figure 11e, an MLT-month coordinate frame with 1 month intervals as shown in Figure 11j, and an MLT-year coordinate frame with 1 year intervals as shown in Figure 11o, respectively. Figure 11 also shows the variations of the four input parameters of the NSW-GDP model and the averaged curves are obtained using data from 1980 to 2015.

Figure 11e shows that the plasmapause shape experiences two weak valleys around 0 h UT and 12 h UT at all MLT sectors, basically consistent with those in Paper 1. Figures 11a-11d indicate that V_{sw} , IMF B_z , and $SYM-H$ have almost no diurnal variation; AE has two weak peaks between ~12 h and ~18 h UT. The differences between Figure 11e and Figure 11a in Paper 1 might be caused by the fact that the plasmapause locations are considered to be equal in the 0 h – 6 h MLT sector in the construction of the basic framework of the model. Also, the diurnal variations of the plasmapause

may be faded through the average of the 36-year data in 1 h UT intervals.

The seasonal variations of the plasmopause locations characterized by equinox valleys and solstice peaks are clearly shown in Figure 11j. Such variations may possibly be caused by the seasonal variations of *SYM-H* shown in Figure 11h since the seasonal variations of V_{SW} , IMF B_z , and *AE* are not so obvious as shown in Figures 11f, 11g and 11i.

The solar cycle variations characterized by solar maximum valleys and solar minimum peaks are well captured by the NSW-GDP model, as shown in Figure 11o. A new feature in Figure 11o is that the solar cycle variations of the plasmopause locations seems to have a time delay relatively to the sunspot number as shown by the dashed curve Figure 11o. This might be caused by the fact that the solar wind conditions and geomagnetic activities are more disturbed during the descending phase of the solar cycle as indicated in Figure 11k-11n and also in the statistics of *Papitashvili et al.* [2000].

4.4. Potential Applications of the NSW-GDP Model

The newly compiled NSW-GDP model is a solar wind driven, dynamic and MLT-continuous plasmopause model. It could be included in dynamic radiation belt and ring current models, such as the Comprehensive Ring Current Model (CRCM) [*Fok et al.*, 2001], the Radiation Belt Environment (RBE) model [*Fok et al.*, 2011], and the Comprehensive Inner Magnetosphere-Ionosphere (CIMI) Model [*Fok et al.*, 2014], instead of the CA1992 (or any of these other) plasmopause models. The NSW-GDP model can also be embedded into the MHD models [e.g., *Raeder et al.*, 2001; *Hu et al.*, 2007; *Pulkkinen et al.*, 2008; *Lu et al.*, 2013] to better characterize the inner boundary of the magnetosphere which is usually set to be at $3 R_E$ in these

MHD models for simplicity.

The NSW-GDP model can also be used to forecast the status of the plasmasphere. If upstream observations of solar wind and IMF parameters and prediction models of *SYM-H* [Cai *et al.*, 2010] and *AE* [Takalo and Timonen, 1997; Luo *et al.*, 2013] are used, the NSW-GDP model can potentially be used to forecast the plasmopause location and shape on the magnetic equatorial plane. Once the plasmopause location is determined, the plasmaspheric plasma density and composition can be predicted through extrapolation of ionospheric density to magnetosphere, such as the method used in the GCPM model [Gallagher *et al.*, 2000].

5. Conclusion

In this paper, the NSW-GDP model is developed based on the largest database to date, containing 49119 plasmopause crossing locations and 3957 plasmopause profiles (corresponding to 48899 plasmopause locations in 1 h MLT intervals), from 18 satellites during 1977 – 2015. Through investigation of the correlations of the plasmopause locations with solar wind parameters and with geomagnetic indices, the 5-min V_{SW} , 5-min B_Z , 5-min *SYM-H*, and 5-min *AE* are chosen as the free, driving parameters of the NSW-GDP model. A time shift in the solar wind parameters and geomagnetic indices relative to derived plasmopause locations is considered in the model development. The driving parameters for the plasmopause model are selected according to the correlation analysis, and the relationships between these parameters and the plasmopause location and shape are constructed by the Levenberg-Marquardt method for nonlinear multiparameter fitting. It is noted that continuous MLT-dependence is embedded in this new model. The limitations of this

model are also discussed. The main results are obtained as follows:

1. The functions of NSW-GDP model are given in equation (14) with the coefficients listed in Table 4.
2. The new model is developed and parameterized by V_{SW} , B_Z , $SYM-H$, and AE . The plasmopause locations are calculated on the magnetic equatorial plane in SM coordinate system.
3. Continuous and smooth MLT-dependence is embedded in this model and controlled mainly by the solar wind driven convection electric field E_{SW} .
4. The NSW-GDP model can well capture the seasonal and solar cycle variations of the plasmopause.
5. This new model improves the forecast accuracy and capability for the global plasmopause, and the best performance of agreements between model results and observations is achieved compared with selected previous models.

In this paper, we have preliminarily developed an initial plasmopause model. Assuming that the plasmasphere is aligned with magnetic field lines and using empirical geomagnetic field models (IGRF and Tsyganenko models), an empirical three-dimensional plasmopause shape can be obtained through field line tracing. The NSW-GDP can potentially be included to current magnetospheric dynamic models and be applied to forecast the dynamic evolution of the plasmasphere. More work should be done in the future to improve the global dynamic plasmopause model, especially for extreme solar wind conditions. Besides, if the time history of the plasmopause and the corotation effect can be embedded into the NSW-GDP model, the performance of the model

can be further improved in the future.

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Tables:

Table 1. Information of Various Empirical Plasmapause Models

Model	Data Source	Year	Parameter ^a	MLT-Dependence
CAA-1992	ISEE-1	1977-1983	Kp	No
MOL-2002	CRRES	1990-1991	Kp	No
OBM-2003	CRRES	1990-1991	Kp , Dst , or AE	Yes
LAR-2007	IMAGE	2001	B_Z and ϕ	No
CHO-2015	THEMIS	2008-2012	B_Z , and AE	No
LIU-2015	THEMIS	2009-2013	$SYM-H$, AL , AU , AE , and Kp	Yes ^b
VER-2015	Cluster	2007-2011	B_Z , BV_{SW} , $d\Phi_{mp}/dt$ ^c , Kp , Dst , or AE	Yes ^d
BAN-2016	CRRES	1990-1991	V_{SW} , BV_{SW} , $d\Phi_{mp}/dt$, Ap , Dst , or AE	Yes ^e
NSW-GDP	Multiple	1977-2015	V_{SW} , B_Z , AE , and $SYM-H$	Yes

^a In this column, “or” indicates that one of the parameters is used in the model, e.g. the OBM-2003 contains three models, VER-2015 contains six models, and “and” indicates that all of the parameters are included in the model.

^b The MLT resolution is 1 h.

^c $d\Phi_{mp}/dt = V_{SW}^{4/3} B_T^{2/3} \sin^{8/3}(\theta_c/2)$, where $B_T = \sqrt{B_Y^2 + B_Z^2}$ is the projection of IMF on the GSM yz plane, $\theta_c = \text{atan}(B_Y, B_Z)$, and B is the IMF magnitude.

^d Just considered three MLT sectors (1 h – 7 h, 7 h – 16 h, and 16 h – 1 h), in each of which the plasmapause location is constant.

^e Only the Ap , Dst , or AE model is MLT-dependent.

Table 2. Fitted Coefficients for the Curves in Figure 3

f_0	f_1	f_2	g_0	g_1	g_2	g_3	g_4	g_5	g_6
2.7883	9.3345	268.6071	1.7569	3.3058	0.7945	3.6678	27.1041	0.1983	0.0010
± 0.0753	± 0.4005	± 12.6382	± 0.1105	± 0.1498	± 0.0871	± 0.0311	± 1.8705	± 0.0925	± 0.0006

Table 3. Contributions of Storm and Substorm Activities to Plasmapause Shrinkage under Different Solar Wind Conditions.

Solar Wind Condition (mV/m)	$ E_{SW} \leq 2.0$	$2.0 < E_{SW} \leq 4.0$	$4.0 < E_{SW} \leq 6.0$	$ E_{SW} > 6.0$
$dL_{PP-Substorm} (R_E)$	1.28	1.20	1.08	1.00
$dL_{PP-Storm} (R_E)$	1.59	1.41	1.31	1.16

Table 4. Fitted Coefficients for the NSW-GDP Model and Corresponding Standard Deviations (σ)

Coefficient	Value	σ	Coefficient	Value	σ	Coefficient	Value	σ
g_{00}	2.051	0.039	g_{52}	143.584	5.011	k_{02}	-0.405	0.0132
g_{01}	28.753	1.938	g_{60}	10.815	0.768	k_{03}	0.0175	0.00093
g_{02}	71.077	3.593	g_{61}	95.789	4.668	k_{04}	-0.045	0.0022
g_{10}	1.627	0.073	h_{00}	1.139	0.0262	k_{05}	-0.005	0.00019
g_{11}	9.065	0.389	h_{01}	-0.0522	0.0050	k_{10}	-0.513	0.0134
g_{12}	259.032	11.952	h_{02}	0.0124	0.0017	k_{11}	0.0165	0.0011
g_{20}	1.317	0.027	h_{10}	0.630	0.0234	k_{12}	0.276	0.0129
g_{21}	34.702	3.286	h_{11}	-0.236	0.0045	k_{13}	-0.0329	0.0024
g_{22}	99.965	6.463	h_{12}	0.0124	0.0015	k_{20}	1.175	0.0456
g_{30}	3.722	0.038	h_{20}	16.656	0.251	k_{21}	-0.075	0.0043
g_{31}	-0.00124	0.00007	h_{21}	0.268	0.0073	k_{22}	-1.825	0.078
g_{40}	15.691	0.443	h_{30}	0.752	0.0227	k_{23}	0.125	0.014
g_{41}	130.797	7.632	h_{31}	-0.301	0.0131	k_{30}	0.75	0.05
g_{42}	179.516	8.620	h_{32}	0.238	0.0147	k_{31}	-0.25	0.01
g_{50}	0.0239	0.003	k_{00}	-0.165	0.0089	k_{32}	4.0	0.25
g_{51}	8.068	0.551	k_{01}	-0.0065	0.0003	k_{33}	2.0	0.10

Figures and Captions:

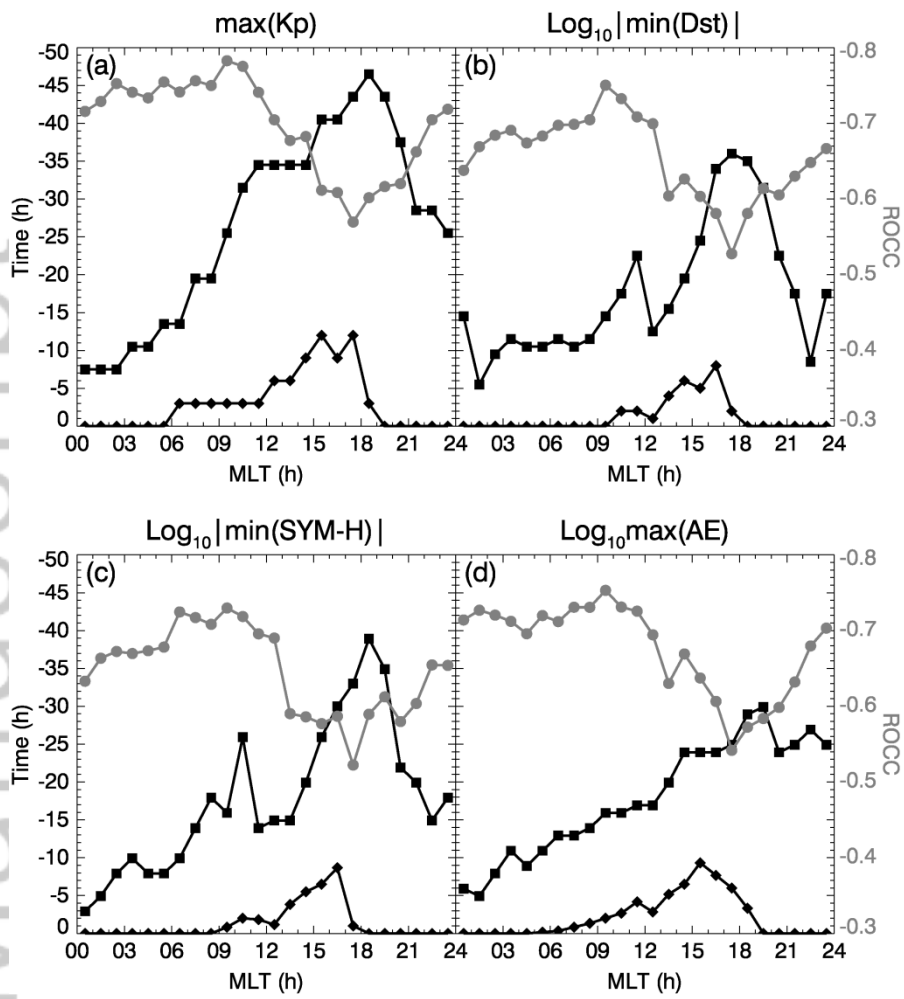


Figure 1. Variations of t_1 (squares) and t_2 (diamonds) for (a) Kp , (b) Dst , (c) $SYM-H$, and (d) AE , respectively, for the strongest correlation with the corresponding ROCC (dots) shown in black.

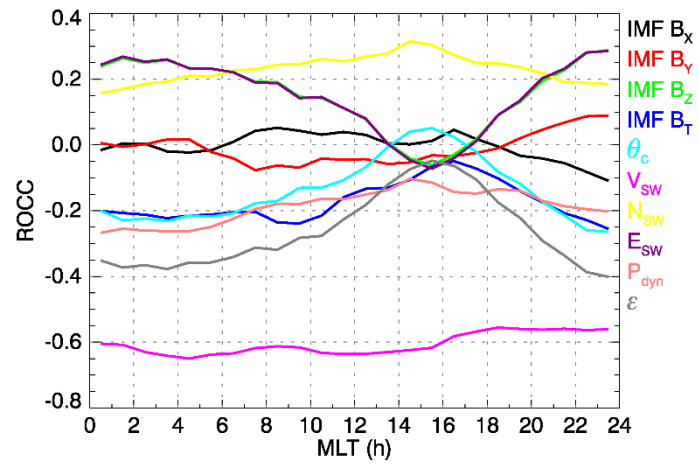


Figure 2. ROCCs between L_{PP} and various solar wind and IMF parameters. Different colors representing different parameters as shown at the right side. Note that the green IMF B_z curve is directly beneath the purple θ_c curve.

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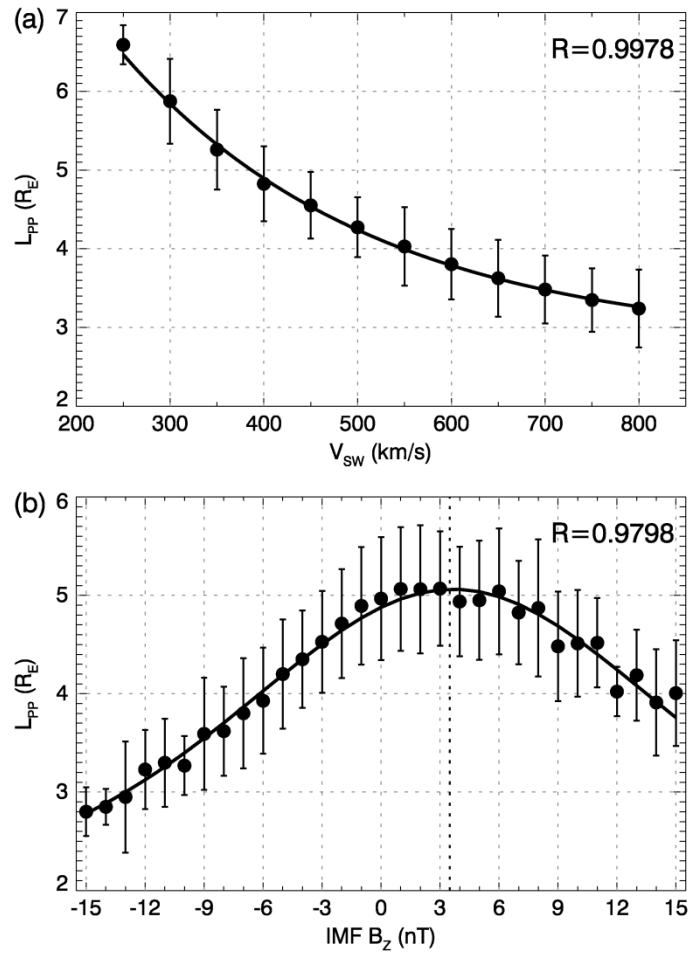


Figure 3. Averaged L_{PP} versus (a) V_{SW} and (b) IMF B_Z , respectively. The black dots represent the averaged L_{PP} in 50 km/s intervals for V_{SW} in the top panel and the averaged L_{PP} in 1 nT intervals for IMF B_Z in the bottom panel with the standard deviations shown by the vertical bars. The curve in (a) represents the best fitting of the dots to an exponentially decreasing function, and the curve in (b) represents the best fitting of the dots to an inverse polynomial function with the peak represented by the vertical dashed line. The correlation coefficients (R) are shown at the top left corners of each panel.

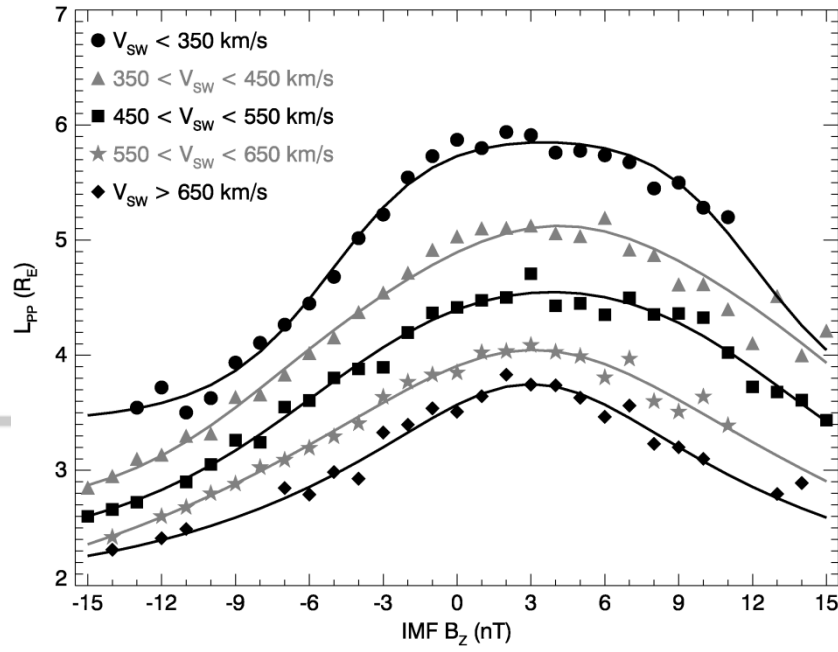


Figure 4. L_{PP} - B_Z variations for different V_{SW} . The dots represent the averaged L_{PP} in 1 nT B_Z intervals. The curves represent the best fittings to equation (4). Different symbols representing different V_{SW} values are shown at the upper left corner.

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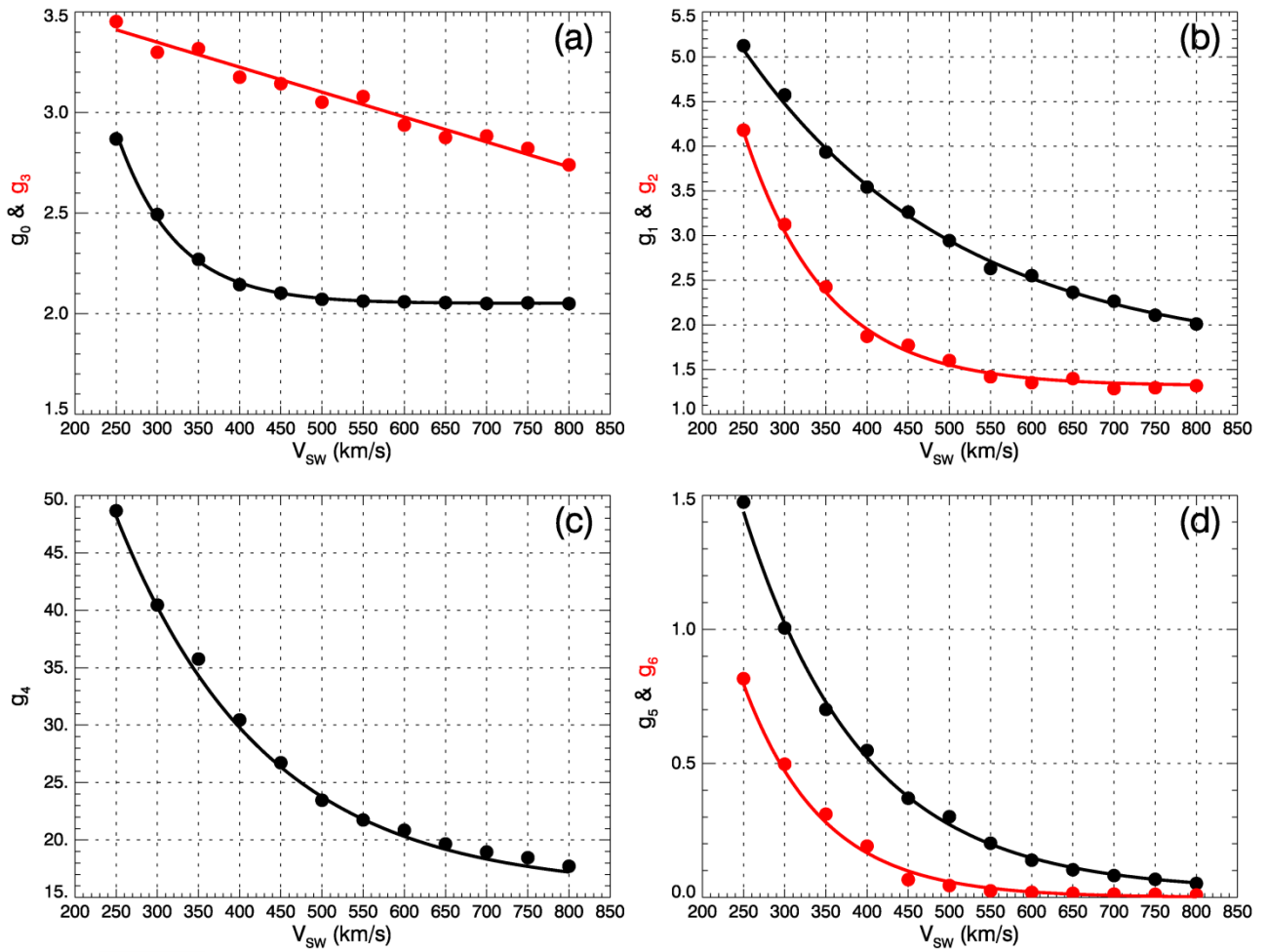


Figure 5. Example of best fittings of the functions in equation (5) for (a) g_0 (black) and g_3 (red), (b) g_1 (black) and g_2 (red), (c) g_4 , and (d) g_5 (black) and g_6 (red). In each panel, the dots represent the values of the parameters fitted using equation (5), and the thick lines represent best fittings of the dots to the equation (6).

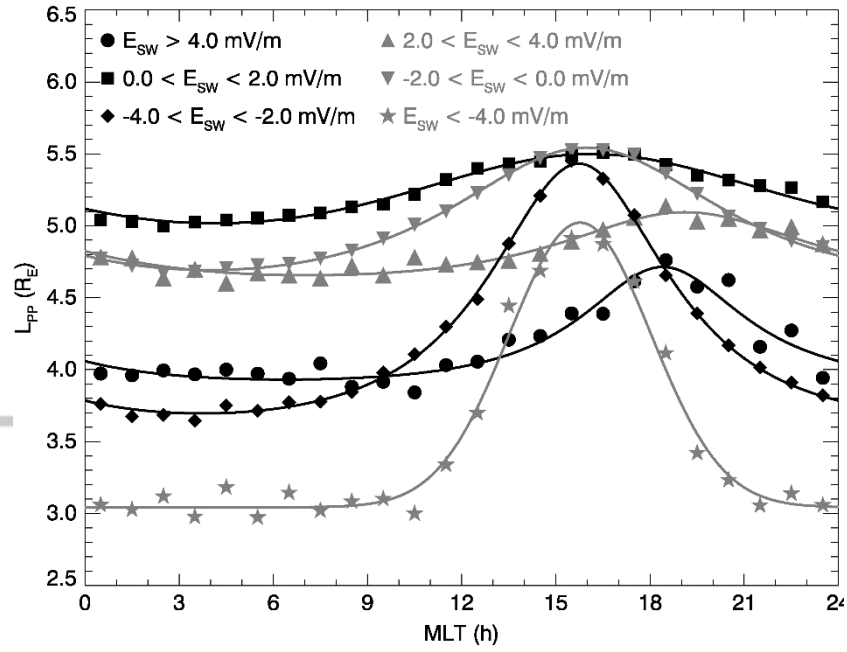


Figure 6. MLT-dependence of the plasmapause locations under different conditions of E_{SW} . The dots represent the averaged L_{PP} in 1 h MLT intervals. The curves represent the best fittings to equation (7). Different symbols representing different E_{SW} values are shown at the upper left corner.

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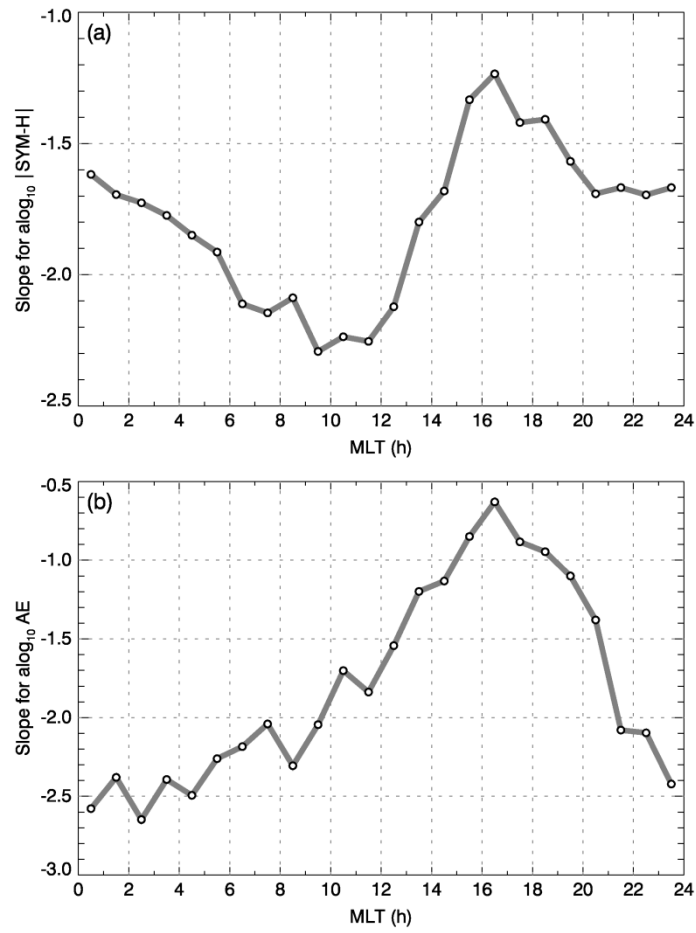


Figure 7. Slopes of the linear fitting functions of L_{PP} to (a) $\log_{10} |SYM-H|$ and (b) $\log_{10} AE$ at different MLT sectors in 1 h intervals, respectively.

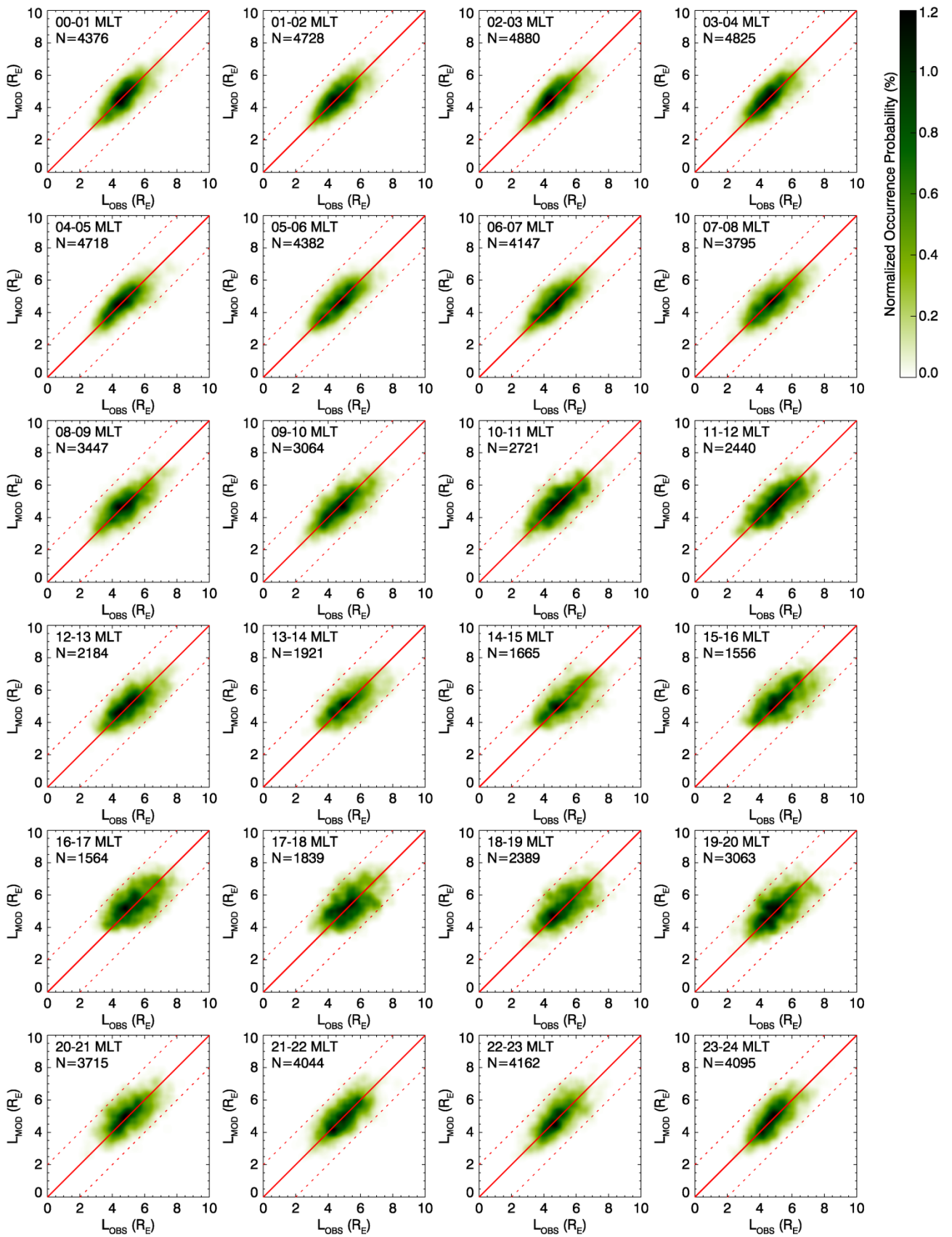


Figure 8. Normalized occurrence probabilities of the L_{MOD} relative to L_{OBS} . The black solid lines indicate $L_{\text{MOD}} = L_{\text{OBS}}$, and the dashed lines are plotted at $L_{\text{MOD}} = L_{\text{OBS}} \pm 2.0$, respectively. The linear scaled color bar denoting the occurrence probabilities are shown at the upper right corner. The numbers of plasmopause locations in each MLT sector are shown at the upper left corners of each panel.

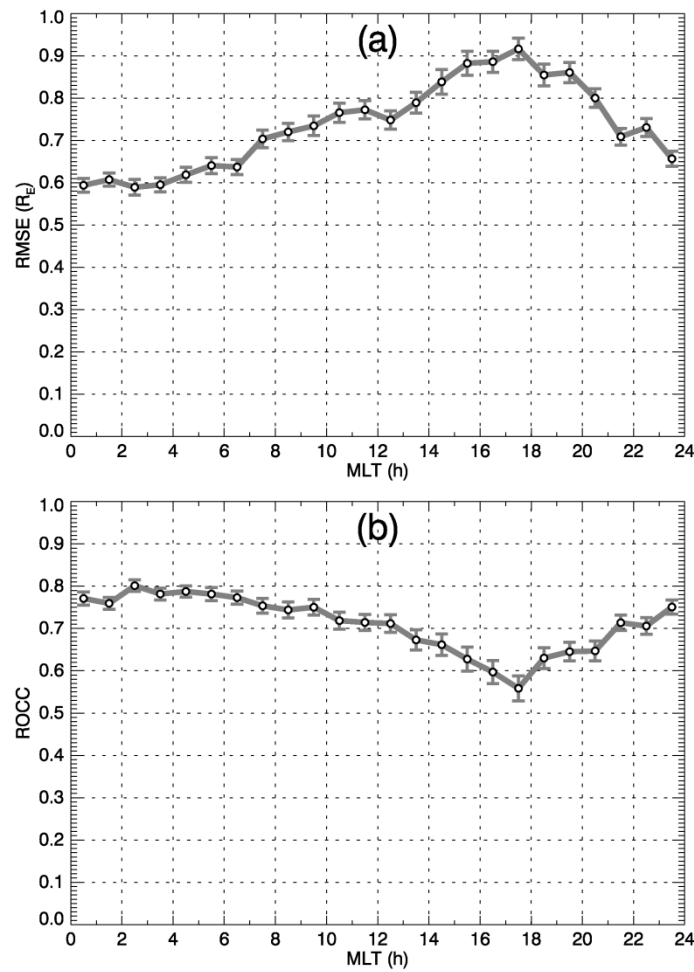


Figure 9. The RMSE and ROCC of the NSW-GDP model versus MLT. The error bars denote the 1σ confidence found using bootstrap resampling.

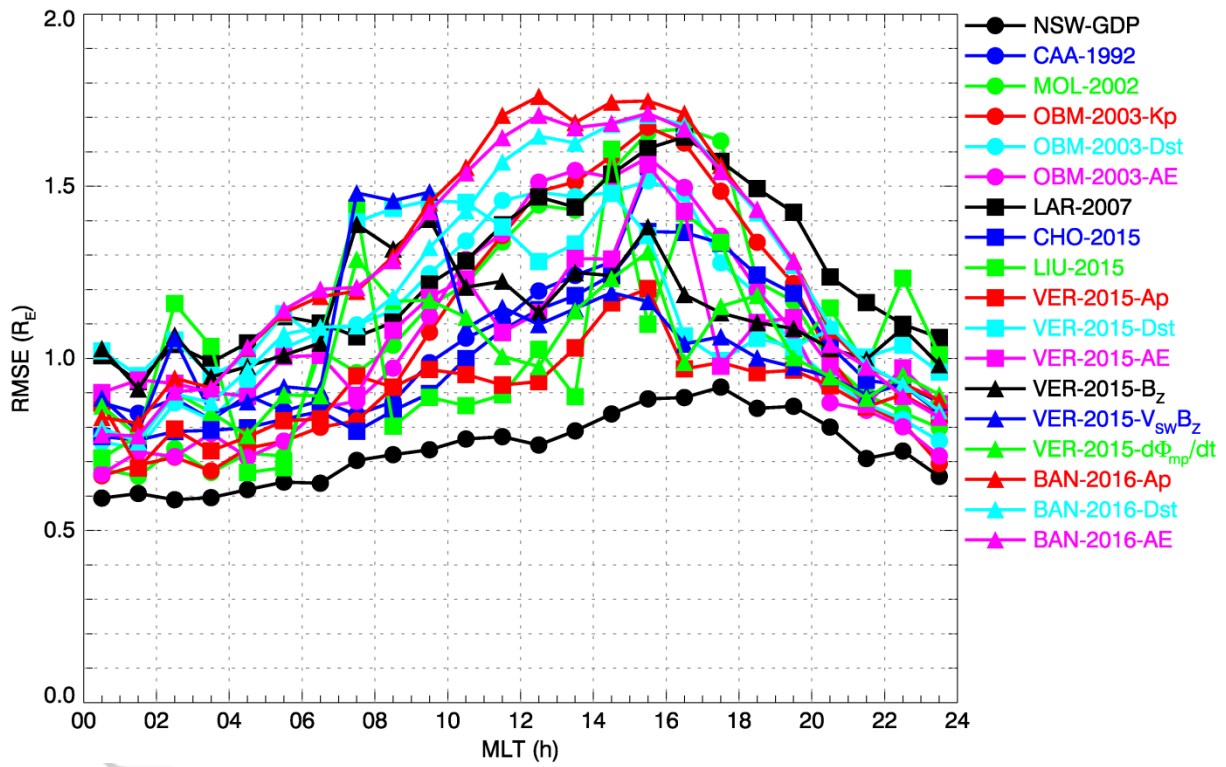


Figure 10. RMSE of various models as a function of MLT. Different colors and symbols representing different models are shown at the right.

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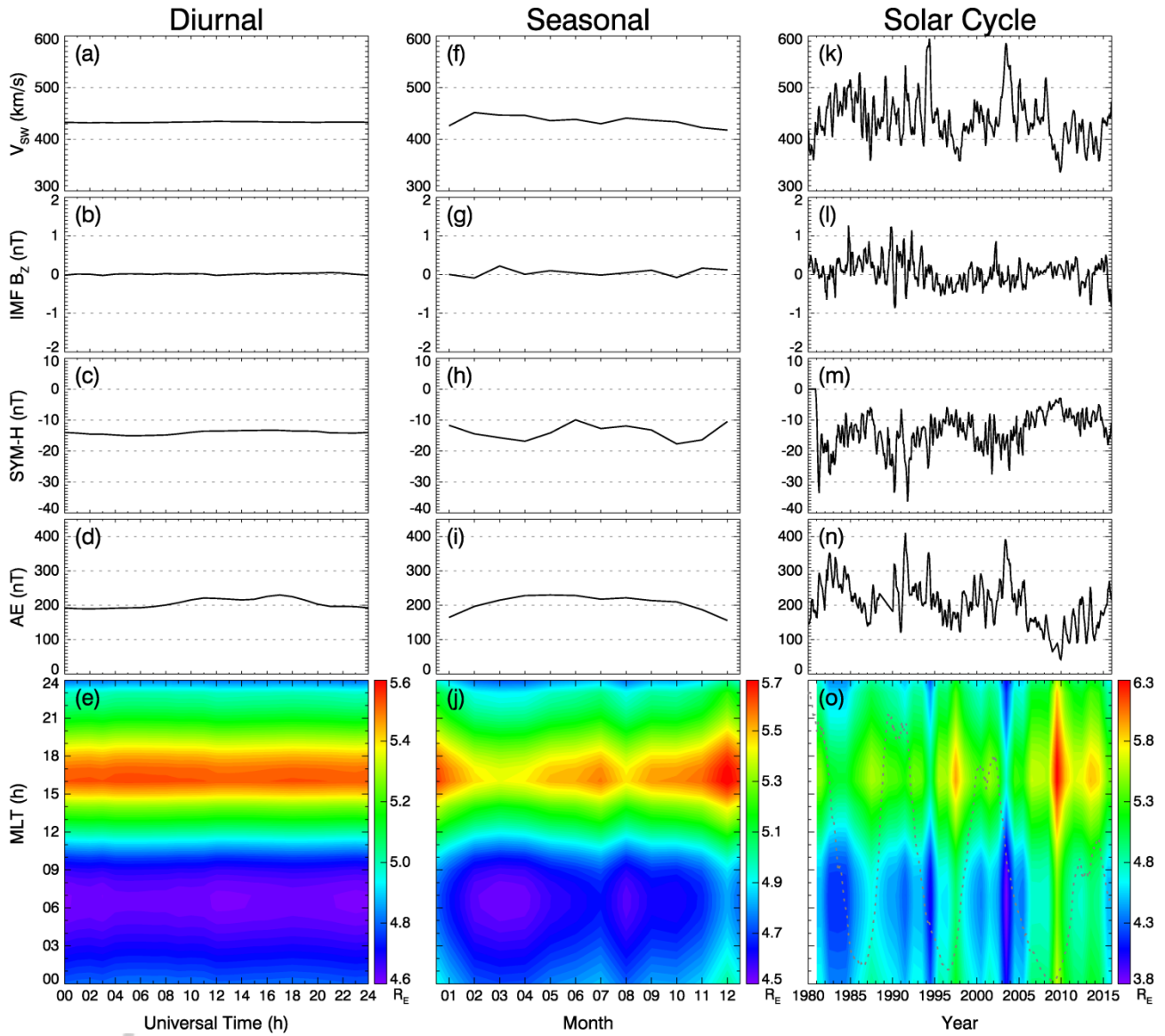
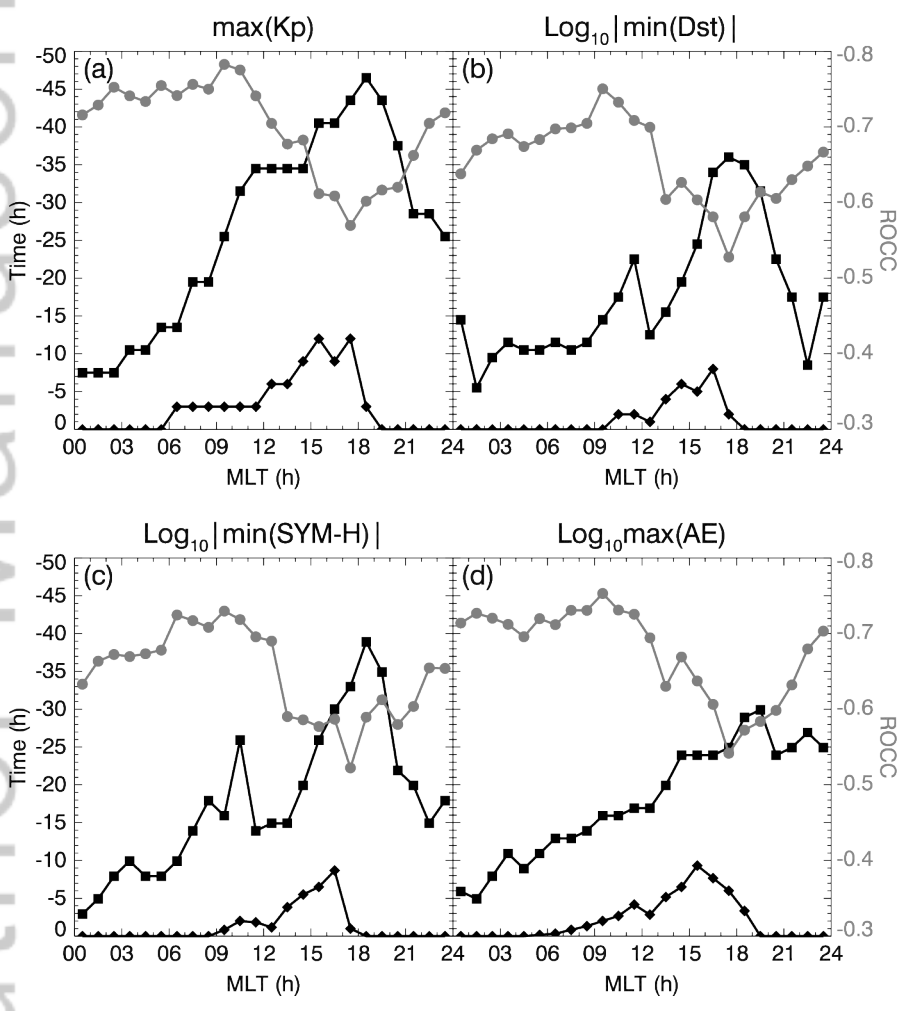


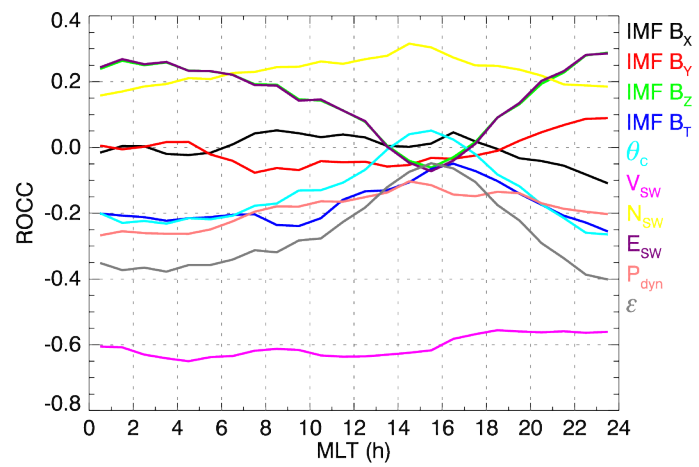
Figure 11. Diurnal (*left column*), seasonal (*middle column*), and solar cycle (*right column*) variations of the plasmopause locations simulated by the NSW-GDP model. From top to bottom in each column, shown are variations of the averaged V_{SW} , IMF B_z , SYM-H, and AE, respectively. From left to right, the averages are calculated hourly, monthly, and monthly, respectively, except for panel (o) where the plasmopause locations are averaged yearly. The grey dashed curve in (o) is the scaled sunspot numbers indicating solar cycles. Note that panels (e), (j) and (o) have different color

bars, which are scaled as shown to the right of each panel.

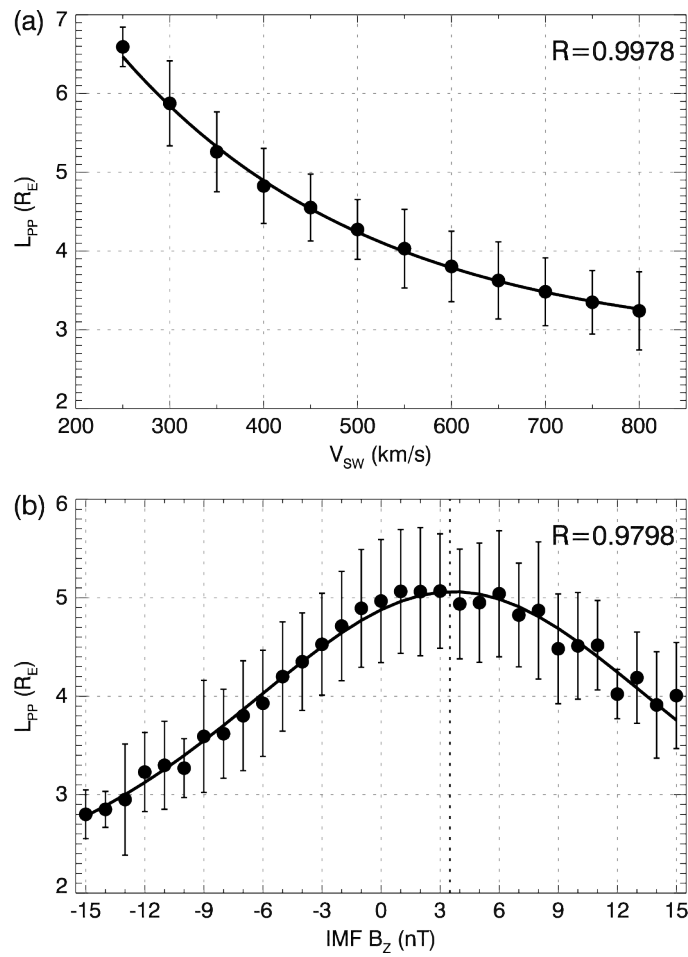
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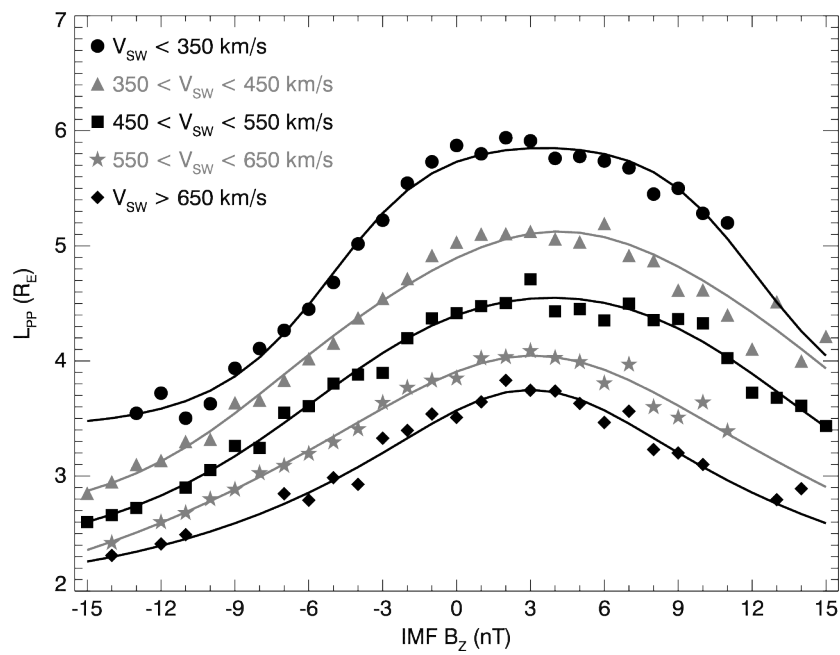
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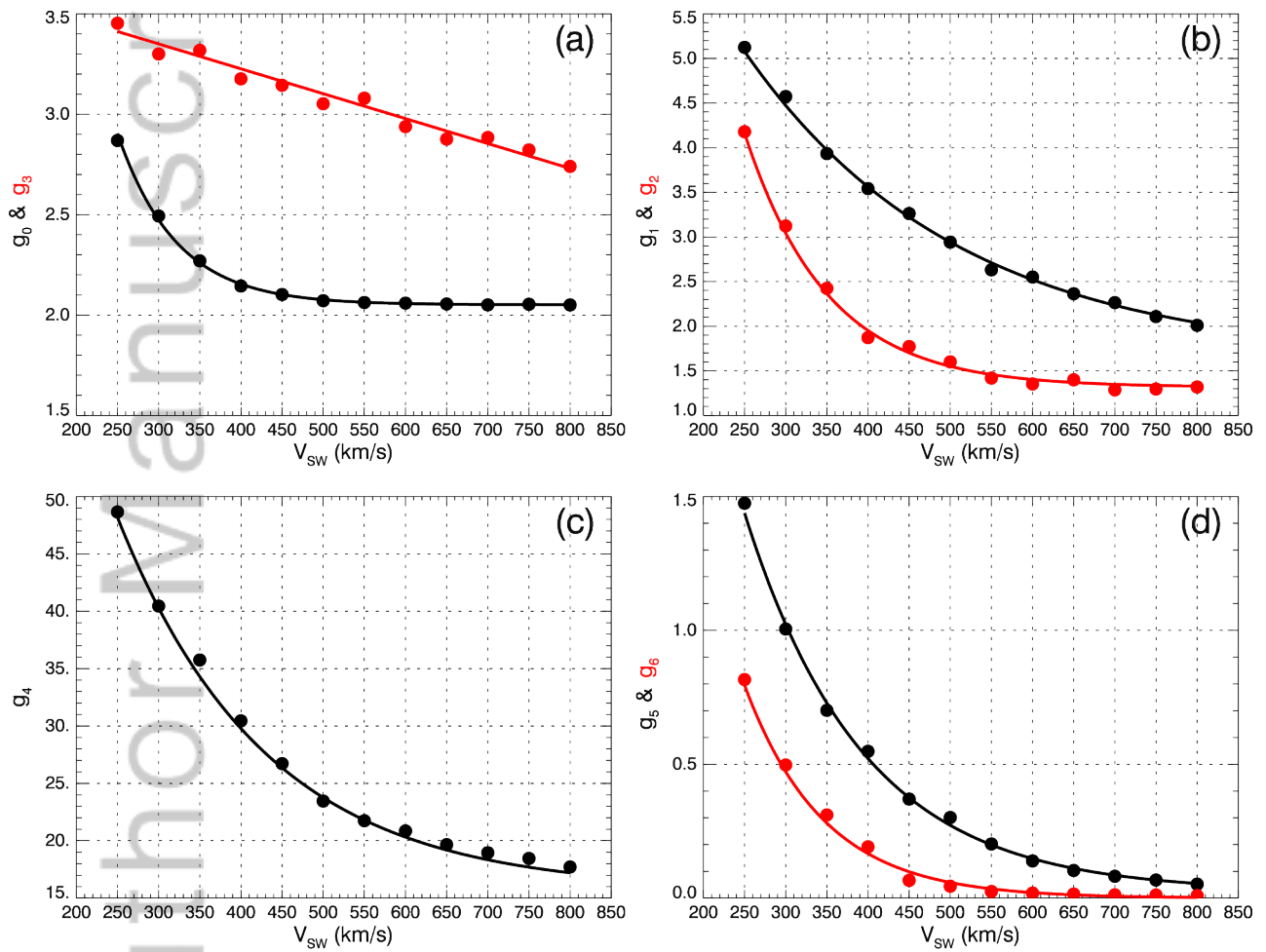
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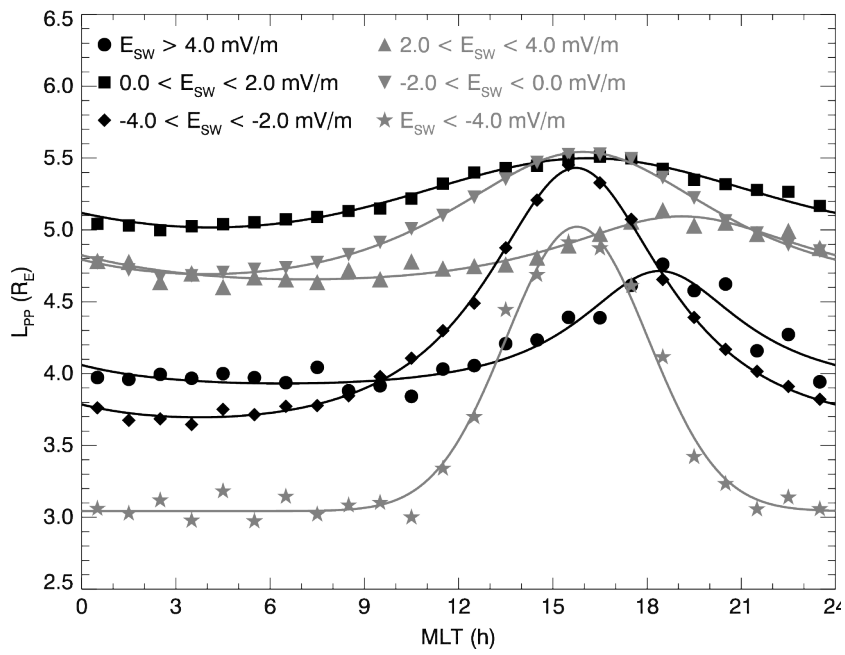
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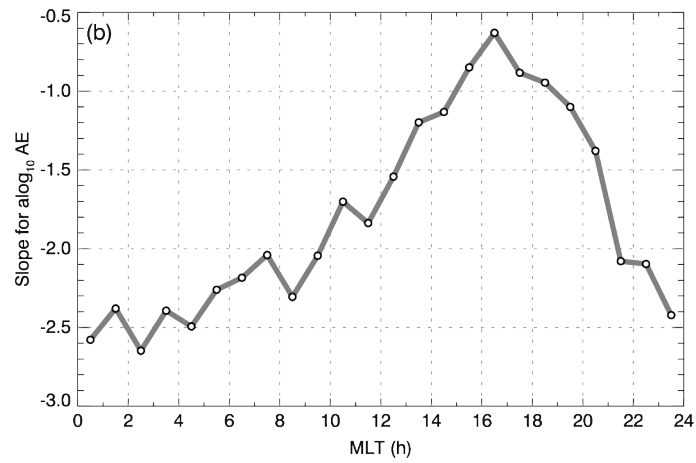
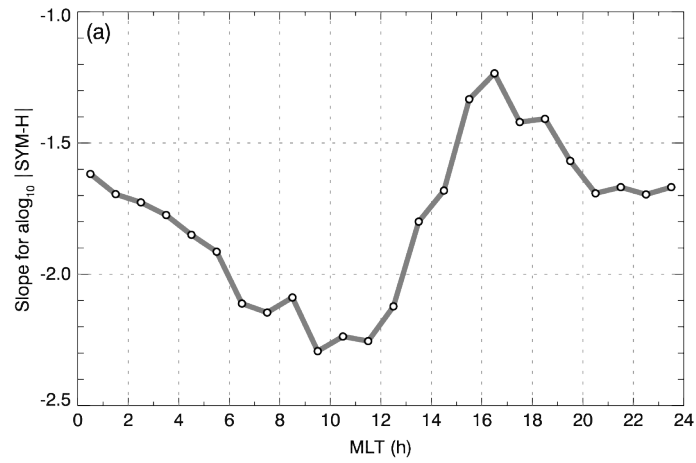
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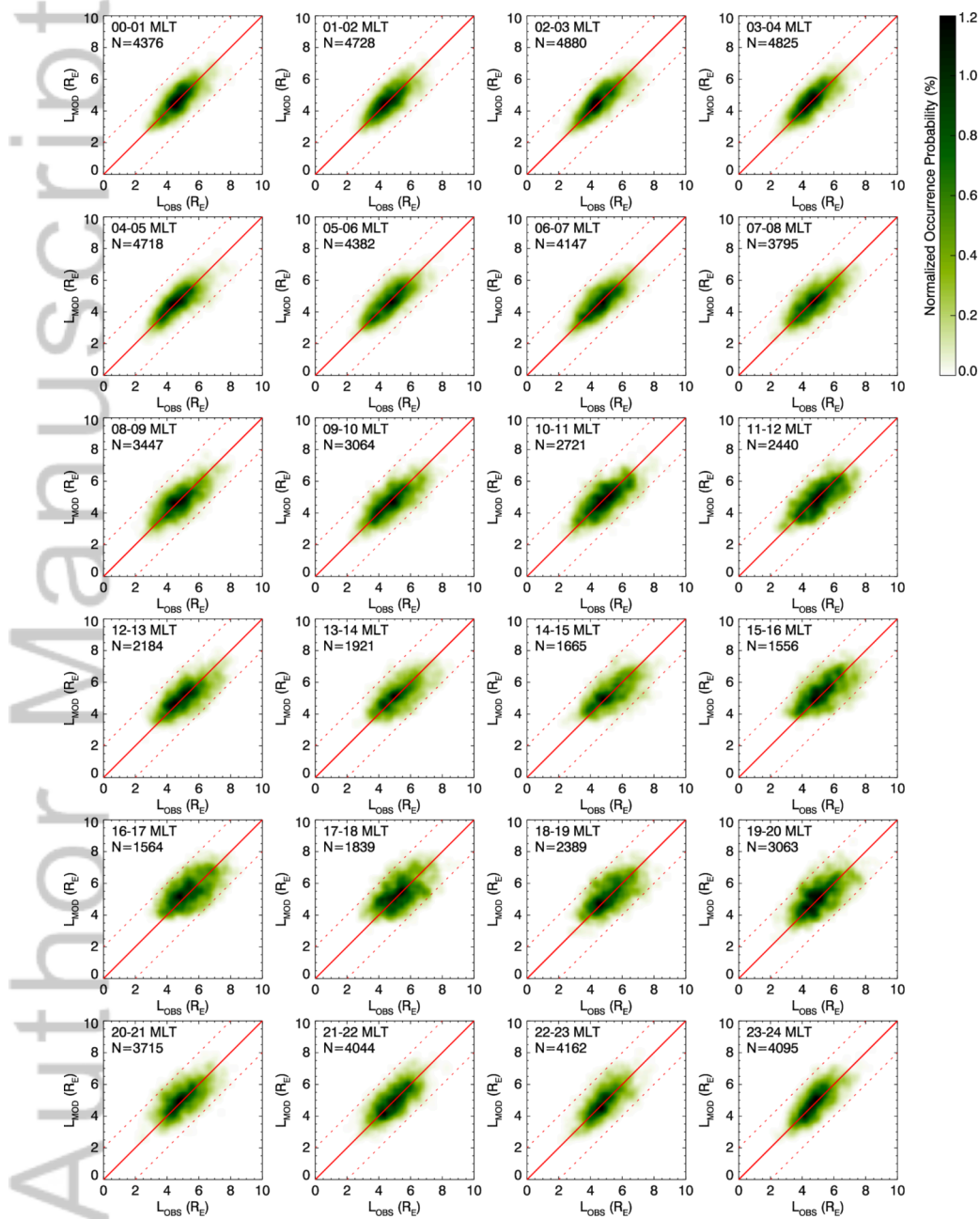
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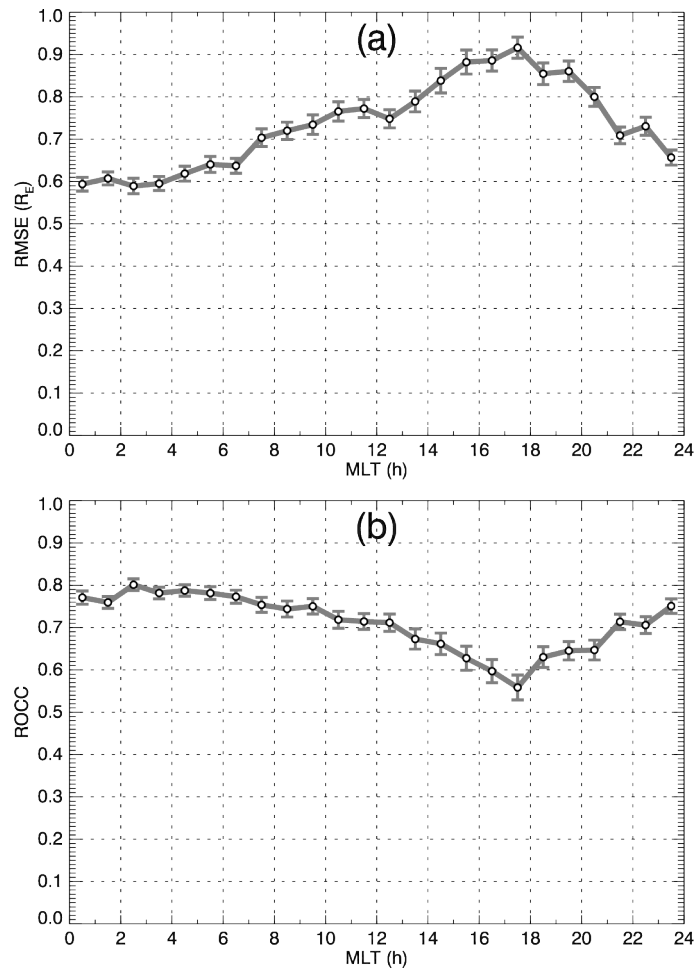
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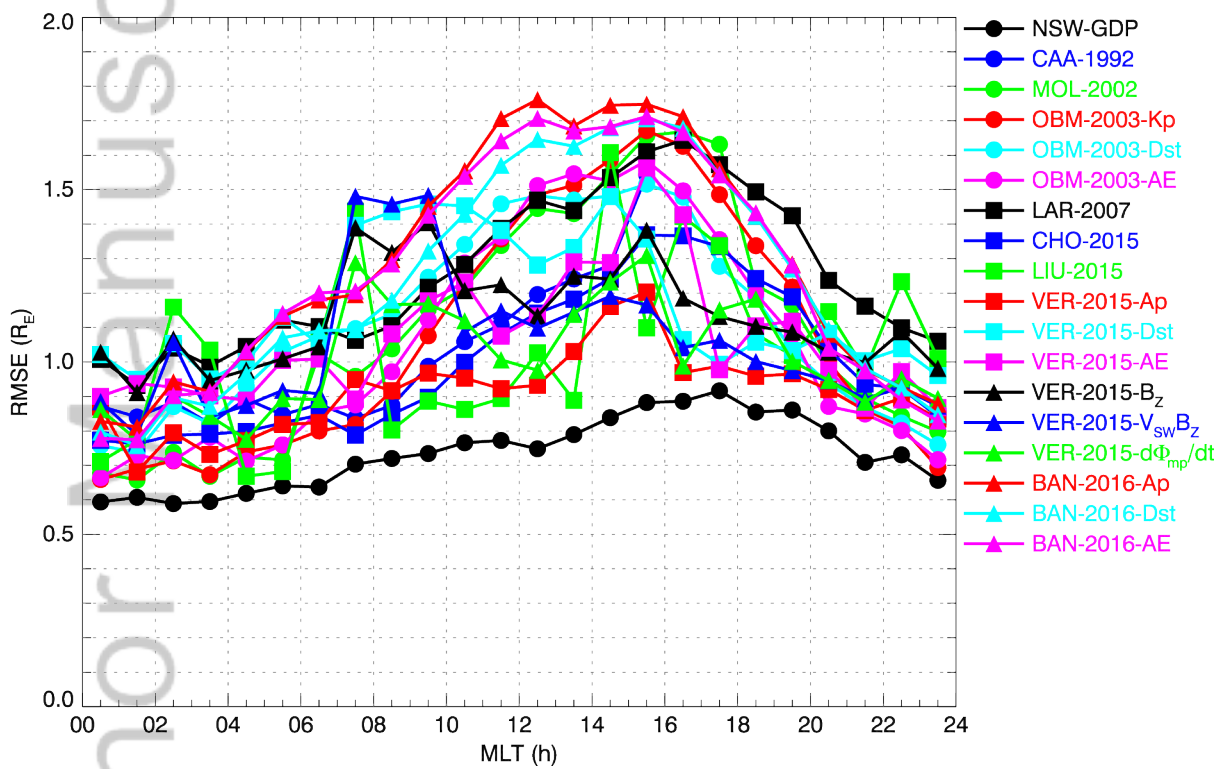
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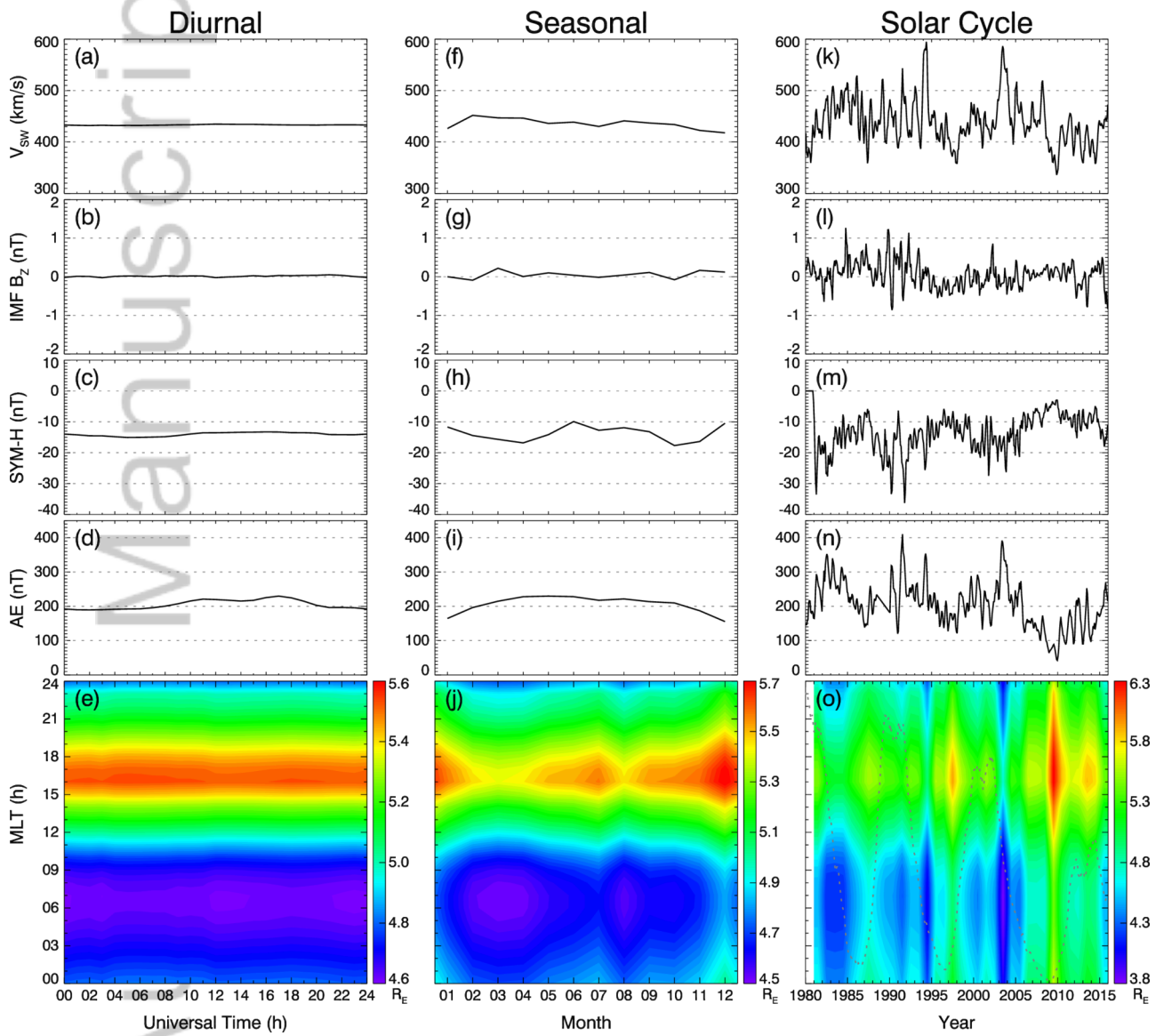
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