Using k-Means Cluster Analysis to find and classify Ion Density Irregularities consistent with Equatorial Plasma Bubbles from Multi-year Formosat-5 Advanced Ionospheric Probe observations

Cornelius Csar Jude Hisole Salinas¹, Loren C. Chang¹, Chi-Kuang Chao¹, Jann-Yenq Liu¹, and Charles C. H. Lin²

¹National Central University ²National Cheng Kung University

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Abstract

This work explores the results of applying Square Euclidean and Correlation k-means cluster analysis on ion density irregularity profiles observed by the Advanced Ionospheric Probe (AIP) onboard the Formosat-5 (F-5) satellite from November 2017 to November 2020. The Square Euclidean cluster analysis yielded separate clusters each for ion density irregularities consistent with Equatorial Plasma Bubbles (EPBs) occurring over the southern low-latitude, northern low-latitude and equator. The Correlation k-means cluster analysis only yielded one cluster characterized by ion density irregularities consistent with EPBs occurring over the equator. Thus, this work suggests that a cluster analysis preferably a Square Euclidean Cluster analysis can be used to find and classify ion density irregularities consistent with EPBs. This work also shows that the F-5/AIP can perform multi-year observations of ion density irregularities at ~710 km altitude and at ~2230 pm local-time that are consistent with irregularities due to EPBs.

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5	Charles C.H. Lin ³
6	¹ Department of Space Science and Engineering, National Central University, Zhongli, Taiwan
7	² Center for Astronautical Physics and Engineering, National Central University, Zhongli,
8	Taiwan
9	³ Department of Earth Science, National Cheng Kung University, Tainan, Taiwan
10	Corresponding author: Loren C. Chang, Department of Space Science and Engineering, National
11	Central University, Zhongli, Taiwan (loren@g.ncu.edu.tw)
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the F-5/AIP can perform multi-year observations of ion density irregularities at ~710 km altitude
and at ~2230 pm local-time that are consistent with irregularities due to EPBs.

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Plain Language Summary

Equatorial plasma bubbles (EPBs) predominantly cause the problems of space-based 25 26 communication and navigation systems over the low-latitudes. These communication and 27 navigation systems involve the use of signals that propagate through our ionosphere. EPBs cause 28 problems in these systems by disturbing the ions along the path of these signals. A well-known 29 way of characterizing the occurrence of these EPBs is to look at the fluctuations or irregularities 30 of ionospheric ion density profiles and determine whether such irregularities are due to EPBs. 31 However, satellite observations have already amassed millions of ion density profiles. Classifying these profiles is a major challenge. This work aims to help tackle this challenge by presenting a 32 33 data-driven approach to classifying these profiles. This work shows the use of k-means cluster analysis on ion density profiles measured by the Advanced Ionospheric Probe onboard the 34 Formosat-5 satellite. This work will show that the approach successfully finds and classifies ion 35 density irregularities that have characteristics consistent with EPBs. 36

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38 Index Terms/ Keywords: Ion Density, Equatorial Plasma Bubbles, Ionosphere, Data Science

39 Key Points:

40	•	K-means	cluster	analysis	is	applied	to	multi-year	ion	density	irre	pularity	profiles
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- K-means cluster analysis finds clusters of ion density irregularities consistent with
 equatorial plasma bubbles.
- Formosat-5's Advanced Ionosphere Probe observes ion density irregularities consistent
 with equatorial plasma bubbles.

45 **1. Introduction**

Ionospheric irregularities cause significant disruptions in satellite communication and 46 navigation systems (Basu et al, 2001). In the equatorial region, equatorial plasma bubbles (EPBs) 47 cause most of these ionospheric irregularities. They form as a result of the nonlinear evolution of 48 49 the generalized Rayleigh-Taylor instability in the night-time ionosphere (Dungey, 1956; Kelley, 50 1989). EPBs grow to be several hundred kilometers in the east-west direction, thousands of 51 kilometers in the north-south direction and also hundreds of kilometers in the vertical direction 52 (Kil, 2015; Yokoyama, 2017). Their overall shape also varies significantly. Thus, observations and 53 identification of EPBs are a challenge.

54 While observing the complete three-dimensional image of an EPB is currently still impossible, we further our knowledge on EPB structure by analyzing when, where and under what conditions 55 they frequently occur. In the past three decades, significant progress has been made in 56 characterizing the occurrence rates of EPBs by analyzing the irregularities observed using satellite 57 58 in-situ ion density measurements. When looking at ion or electron density profiles, EPBs manifest as sudden drops in density values. These analyses involve grouping the ion density profiles often 59 in terms of EPB parameters. One of the most commonly used parameters is the depth of depletion 60 $\Delta N/N_0$ where N_0 is a background ion density and ΔN is the perturbation. In the early 2000s, the 61 first global climatology of EPB occurrence rates was formed by applying this method on more 62 63 than a decade of measurements from 6 Defense Meteorological Satellite Program (DMSP) satellites (Huang et al, 2001; Burke et al, 2004; Gentile et al, 2006). DMSP satellites are in a 64 circular sun-synchronous polar orbit at an altitude of ~840 km and they all cross the magnetic 65 66 equator post-sunset. With this orbit, the Special Sensor-Ions, Electrons and Scintillation (SSIES) 67 instruments onboard consistently measure night-time in-situ ion density latitude profiles with

specific local-time coverages ranging between 1900 and 2100 pm local-time. These provided EPB occurrence rates as a function of longitude and month. It showed that EPBs inducing large plasma depletions occur most frequently during equinox and that in this season, most of the EPBs occur over the American, Atlantic and African sectors. They also showed that EPBs generally follow the magnetic declination angle. These are all consistent with theoretical studies (Tsunoda, 1985).

73 The same approach was utilized on measurements by the Republic of China Satellite -174 (ROCSAT-1) as well as the Communication Navigation Outage Forecast System (C/NOFS) 75 satellite. ROCSAT-1 is in a circular orbit at an altitude of 600 km and an inclination of 35 degrees. 76 With this orbit, ROCSAT-1's ascending node regressed westward at a rate of ~7 degrees per day. 77 With this nodal regression, the local-time as well as the magnetic latitude of measurements performed by the Ionospheric Plasma and Electrodynamics Instrument (IPEI) changed 78 79 significantly each day (Yeh et al, 1999; Chen et al, 2001; Su et al, 2001; Le et al, 2003). Evening local-times were measured around 3 to 4 times a day. While this reduced the number of 80 measurements at night compared to DMSP, this enabled the examination of the magnetic latitude 81 dependence of EPBs. Burke et al (2004) analyzed ROCSAT-1 IPEI measurements in March and 82 April of 2000 and 2002. They showed that EPBs occur more frequently around the magnetic 83 84 equator.

C/NOFS used an Ion Velocity Meter (IVM) to measure ion density. It was in an elliptical orbit with a 13-degree inclination. At the time of launch in 2008, its perigee was around 400 km and its apogee was around 850 km. The precession of C/NOFS allowed it to attain full local-time coverage in 2 months. Heelis et al (2010) also grouped the profiles in terms of depth of plasma depletion to determine the seasonal and local-time-dependencies of EPB occurrence rates. They 90 found that EPBs occur most frequently post-midnight and they suggested that this may be related91 to the seeding of EPBs in the bottom-side ionosphere.

More recent studies tried to actually isolate the individual EPBs along an ion density profile. Smith and Heelis (2017) employed edge detection on C/NOFS ion density profiles to isolate individual plasma bubbles along an ion density profile. This determined the occurrence rates of different scales of plasma bubbles. They found that most bubbles have a depletion length of around 200 km. Wan et al (2018) processed SWARM data to isolate the depletions and then the amplitude of these depletions. This determined the occurrence rate of different EPB intensities.

98 This work builds off of these previous studies in two ways. One, we present multi-year observations of ion density irregularities consistent with EPBs from a new space-based 99 observational platform, the Advanced Ionospheric Probe (AIP) onboard the Formosat-5 (F-5). 100 Two, we present a new method of finding and classifying ion density irregularity profiles 101 consistent with EPBs. As shown by the aforementioned studies, most methods of EPB detection 102 frequently involve grouping ion density profiles in terms of hyper-parameters such as ion density 103 dip magnitude, depletion edge criteria or length of depletion. The chosen values of these 104 parameters are based on looking at just a few ion density profiles. The need to group the profiles 105 106 actually calls for classification algorithms. Classification algorithms offer a more data-driven approach compared to choosing hyper parameters as well as choosing the values for these 107 parameters by only looking at a few ion density profiles. This work does exactly this by utilizing 108 k-means cluster analysis to group ion density irregularity profiles. The clusters formed by the 109 algorithm will then be characterized in terms of when (month) and where (in longitude and 110 latitude) the irregularities occur. These are the first reported multi-year F-5/AIP observations on 111

ion density irregularities consistent with EPBs. To the best of the authors knowledge, this is alsothe first use of cluster analysis for the purpose of ion density irregularity classification.

114 2. Methodology

This work utilizes F-5/AIP night-time ion density profiles from November 2017 to 115 November 2020. F-5 was launched into a repeating sun-synchronous orbit (orbital inclination of 116 117 98.28 degrees) at an altitude of 720 km on August 25, 2017 (Lin et al, 2017; Chao et al, 2020). It orbits along the 1030-2230 LT sectors. This orbit allows AIP to globally sample the 2230 local-118 119 time in just 2 days. Figure 1A and 1B show this sampling. Figure 1A shows the sampling track of 120 AIP for one day while figure 1B shows the sampling track for two days. The geographic latitudinal 121 coverage of AIP, on average, is between -35S and 65N. Since this work is focused on EPBs, we only utilize data between 30S and 30N. AIP utilizes an ion trap sensor to measure ion concentration 122 (Lin et al, 2017). Its sampling rate is up to 8192 Hz which can resolve ion density structures as 123 small as 7.4 m. However, this work utilizes data that is the median of ion density measurements in 124 1 second. This yields orbital profiles with a spatial resolution of around 0.05 degrees or ~5 km. 125 There are approximately 7 orbital profiles per day. To further minimize data-gaps, these profiles 126 are interpolated into 0.5-degree latitudinal resolution. 127

To isolate the irregularities in a given ion density profile and to determine their amplitudes, we first filter the ion density profiles using a 5th-order Butterworth band-pass filter that cuts off fluctuations with horizontal wavelengths less than 200 km and greater than 500 km between 40S and 40N. These are well within the range of EPB dimensions (Kil, 2015; Yokoyama, 2017). Then, to get the amplitudes, we take the square-root of the square of this filtered form. Figure 1C shows a sample profile, its filtered form and the amplitude profile. The strongest irregularities in this sample profile are clearly between 10N and 20N. The filtered form of the profile effectively removes the background ion density profile but retains the strongest irregularities. Note though that this approach isn't meant to mimic the exact dimensions of the unfiltered irregularities. This approach is just meant to provide an amplitude profile whose regions of highest amplitudes coincide with the strongest irregularities. These amplitude profiles are then subject to the cluster analysis.

- 140 K-means cluster analysis is an unsupervised clustering method that groups data in terms of
 141 their similarities (Wu, 2012). The general algorithm of a k-means cluster analysis is as follows:
- 142 Step 1: Given a set of data points, randomly choose N number of initial cluster centers.
- 143 Step 2: Calculate the difference between each data point and the initial cluster centers and144 determine which cluster each data point is closer to.
- 145 Step 3: After all data points are classified, calculate the average of each cluster and set this146 average as the new cluster center.

147 Step 4: Repeat steps 2 and 3 until the cluster centers no longer change.

This work uses MATLAB's built-in k-means cluster analysis function to perform an 8cluster k-means cluster analysis utilizing the square Euclidean distance and the correlation distance similarity metric. This means that the analysis will yield 8 numbers clusters and the square Euclidean distance and the correlation distance are our measures for the differences of the data points. For a dataset made of profiles (e.g. ion density profiles), a square Euclidean distance metric will group profiles whose element-by-element differences are minimal. If *x* and *y* are profiles, the square Euclidean metric is defined as:

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$$||x - y|| = \sum_{i=1}^{d} x_i - y_i$$
 (1)

156 where d is the total number of elements in a profile, x_i is element number i in profile x and y_i is element number *i* in profile *y*. On the other hand, a correlation distance similarity metric will group 157 158 profiles with the highest element-by-element correlation. The standard statistical definition of correlation is used. With regards to the number of clusters, previous studies grouped the profiles 159 160 into 4 groups depending on their depth of plasma depletion. To find the optimal number of groups for a cluster analysis, we look at the average of the sum of data point-to-cluster center distance 161 162 (PtoD distance hereafter) for all groups (Wu, 2012). The lower the PtoD is in a group, the higher the similarity of the data-points in that group. The best number of clusters is the number such that 163 decreasing this number significantly reduces the average PtoD while increasing this number 164 minimally changes the average PtoD. When plotting the average PtoD as a function of number of 165 166 clusters, the value for the best number of clusters is the 'knee' of this plot. We found that 8 clusters satisfied these conditions. Hence, we chose this as the number of clusters. Finally, for each group, 167 we tabulate how many profiles are found on each month for all years from 2017 till 2020 as well 168 169 as for a given longitude bin. This work utilizes a 30-degree longitude bin. This allows us to see the occurrence frequencies of each group as a function of longitudinal sector and month. Previous 170 171 studies calculated occurrence rates as the ratio of the number of profiles for a given group and the 172 total number of possible profiles. However, we found that for a grid of month and longitude, the 173 total number of possible profiles is uneven. This is shown in figure 1D. Hence, we don't use this 174 definition because the uneven distribution will cause biases.

175 **3. Results**

Figure 2 shows the results of the Square Euclidean Cluster Analysis. Figure 2 shows 8 panels with each panel containing two plots. Each panel corresponds to a cluster. The left plot of each panel shows the average of all profiles under the given cluster as a function of geographic 179 latitude. The errorbars are the standard deviation of the profiles. Note that the y-axis for the left 180 plots differ per panel. The right plot of each panel shows the number of profiles found in each 181 month and longitude bin. This will be called the occurrence frequency plot. For example, the 182 occurrence frequency plot of cluster 1 in figure 2 shows that of all profiles for all years, there are 183 only 5 profiles of cluster 1 found in November and over geographic longitude -60°. Note that we 184 don't separate the profiles in terms of year. Hence, this right plot actually shows the seasonality of 185 the occurrence frequency for each cluster.

186 Apart from the geographic location, amplitude and seasonality of these irregularities, it is 187 also important to characterize the magnetic inclination angles over these profiles because this also 188 helps determine whether the irregularity within these profiles may be due to EPBs (Tsunoda, 1985). One way to do this is show a plot of the magnetic inclination angles as a function of latitude 189 190 and longitude. Then, for each cluster, we first determine the latitudinal location of the peak amplitude and then also determine the most frequent longitudinal location of the cluster's profiles. 191 We then check what the magnetic inclination angle is over this latitudinal and longitudinal 192 193 location. This work takes a different approach. Instead of showing a separate plot for the magnetic inclination angles, we include the information on left plot of each panel. To do this, first, for each 194 195 ion density profile, we use MATLAB's built-in IGRF model to determine the magnetic inclination 196 angles at each data point in the profile. Thus, for each ion density profile, there is a corresponding magnetic inclination angle profile. Then, from this magnetic inclination angle profile, we note the 197 average geographic latitude of magnetic inclination angles -15°, 0° and 15°. We only look at these 198 magnetic inclination angles because we are only interested in knowing the magnetic low-latitude's 199 geographic location. Finally, for each cluster, we average the geographic latitudes of magnetic 200 201 inclination angles -15°, 0° and 15°. The red line in each panel's left plot is situated over the average

202 geographic latitude of magnetic inclination angle 0° for the given cluster. The green line to the left 203 (right) of the red line in each left plot is situated over the average geographic latitude of magnetic 204 inclination angle -15° (15°) for the given cluster. Plotting these lines over the cluster's average 205 profile quickly shows whether the ion irregularity profiles in the cluster have peaks found within 206 the magnetic low latitude. If they are, this would be additional evidence that most of the ion density 207 irregularities in the cluster may be attributed to EPBs.

208 For all clusters found using the Square Euclidean cluster analysis, it will be shown that the 209 errorbars are small indicating that the number of profiles deviating from each cluster's average 210 latitudinal variation are minimal. This suggests that most of the profiles in the clusters have 211 magnitudes and latitudinal variations consistent with the average profiles. Figure 2's panel A shows the Square Euclidean cluster-1 irregularities (Sq-Cluster-1 irregularities hereafter and other 212 213 Square Euclidean clusters will follow this labeling). The average profile plot shows that cluster-1 214 irregularities have peak amplitudes of 80 units (units = 10^3 #/cc) between geographic latitudes 10S 215 and 10N. The occurrence frequency plot of Sq-Cluster-1 shows that Sq-Cluster-1 irregularities are 216 mostly found in February and November over the American sector. The green and red lines indicate that the geographic latitude of the peak amplitudes in Sq-Cluster-1 irregularities are within 217 218 the geomagnetic low-latitudes.

Figure 2's panel B shows Sq-Cluster-2 irregularities. Its average profile plot shows that Sq-Cluster-2 irregularities have peak amplitudes of 40 units between latitudes 5S and 5N. The occurrence frequency plot of Sq-Cluster-2 shows that Sq-Cluster-2 irregularities are mostly found in February and in November also over the American sector. The green and red lines indicate that the geographic latitude of the peak amplitudes are within the geomagnetic low-latitudes. Figure 2's panel C shows Sq-Cluster-3 irregularities. Its average profile plot shows Sq-Cluster-3 irregularities have peak amplitudes of 50 units between latitudes 20S and 5S. The occurrence frequency plot of Sq-Cluster-3 shows that Sq-Cluster-3 irregularities are mostly found in February and November over the complete American sector. The green and red lines indicate that the geographic latitude of the peak amplitudes are partially within the geomagnetic lowlatitudes.

Figure 2's panel D shows Sq-Cluster-4 irregularities. Its average profile plot shows Sq-Cluster-3 irregularities have peak amplitudes of 50 units between latitudes 5N and 20N. The occurrence frequency plot of Sq-Cluster-4 shows that Sq-Cluster-4 irregularities are mostly found in March and in November over the Atlantic sector. The green and red lines indicate that the geographic latitude of the peak amplitudes are within the geomagnetic low-latitudes.

Figure 2's panel E shows Sq-Cluster-5 irregularities. Its average profile plot shows Sq-Cluster-3 irregularities have peak amplitudes of 20 units between latitudes 20S and the equator. The location of the peak amplitudes is similar to that of Sq-Cluster-3 but the amplitudes are weaker. The occurrence frequency plot of Sq-Cluster-5 shows that Sq-Cluster-5 irregularities are mostly found from November to April over all longitudes except the Atlantic to Indian sector. The green and red lines indicate that the geographic latitude of the peak amplitudes are within the geomagnetic low-latitudes.

Figure 2's panel F shows Sq-Cluster-6 irregularities. Its average profile plot shows Sq-Cluster-6 irregularities have peak amplitudes of 10 units over latitudes 30S and 10S. The occurrence frequency plot of Sq-Cluster-6 shows that over the East Pacific and American sector, Sq-Cluster-6 profiles are mostly found between October and April while over the West Pacific, Sq-Cluster-6 profiles are found throughout the year with peak occurrence in June. The green and red lines indicate that the geographic latitude of the peak amplitudes are very far from thegeomagnetic low-latitudes.

Figure 2's panel G shows Sq-Cluster-7 irregularities. Its average profile plot shows Sq-249 Cluster-7 irregularities have peak amplitudes of 10 units over latitudes 5S and 20N. The occurrence 250 251 frequency plot of Sq-Cluster-7 shows that Sq-Cluster-7 irregularities are mostly found in March 252 and September over all longitudes except the Pacific and American sectors. Over the American 253 and Atlantic sectors, Sq-Cluster-7 irregularities are mostly found in April and June. Over the East 254 Pacific and American sectors, Sq-Cluster-7 irregularities are mostly found in May and August. 255 The green and red lines indicate that the geographic latitude of the peak amplitudes are partially 256 over the geomagnetic low-latitudes.

Figure 2's panel H shows Sq-Cluster-8 irregularities. Its average profile plot shows Sq-Cluster-8 irregularities have minimal latitudinal variation with magnitudes of around 2 units throughout. The occurrence frequency plot of Sq-Cluster-8 shows that Sq-Cluster-8 irregularities are mostly found between March and September over the American and Indian sectors.

The occurrence frequency plots of Sq-Cluster-1, Sq-Cluster-2, Sq-Cluster-3 and Sq-261 262 Cluster-4 show that the ion density irregularities in these clusters frequently occur in months and longitudinal sectors when and where EPBs frequently happen (Huang et al, 2002; Gentile et al, 263 2006). These are months and sectors when the magnetic field is aligned with the solar terminator 264 enhancing EPB rates (Tsunoda, 1985). In addition, the latitudinal location of these irregularities 265 follow the location of the magnetic low-latitudes. This dependency with the magnetic equator is 266 also consistent with how irregularities due to EPBs behave (Sultan, 1996). Thus, the ion density 267 irregularities comprising Sq-Cluster-1, Sq-Cluster-2, Sq-Cluster-3 and Sq-Cluster-4 may mostly 268 comprise of irregularities associated with EPBs. 269

270 Figure 3 shows the results of the Correlation Cluster Analysis. The figure is formatted in the same way as figure 2. Figure 3's panel A shows the Correlation cluster-1 irregularities (Corr-271 Cluster-1 irregularities hereafter and other Correlation clusters will follow this labeling). The 272 273 average profile plot shows that Corr-Cluster-1 irregularities have peak amplitudes of 15 units between geographic latitudes 10S and 10N. The occurrence frequency plot of Corr-Cluster-1 274 275 shows that Corr-Cluster-1 irregularities are mostly found over the East Pacific, American and 276 Asian sector between February and April as well as between August and November. The green 277 and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-1 irregularities are mostly within the geomagnetic low-latitudes. 278

Figure 3's panel B shows Corr-Cluster-2 irregularities. The average profile plot shows that Corr-Cluster-2 irregularities have peak amplitudes of 12 units between latitudes 10S and 5N. The occurrence frequency plot of Corr-Cluster-2 shows that Corr-Cluster-2 irregularities are mostly found over the Asian, Pacific and the American sectors during equinox seasons. The green and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-2 irregularities are not within the geomagnetic low-latitudes.

Figure 3's panel C shows Corr-Cluster-3 irregularities. The average profile plot shows that Corr-Cluster-3 irregularities have peak amplitudes of 9 units between latitudes 20S and 5S. The occurrence frequency plot of Corr-Cluster-3 shows that over the Asian sector, Corr-Cluster-3 profiles are found throughout the year with the most occurrences in June. It also shows that over the American sector, Corr-Cluster-3 profiles are found between October and March. The green and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-3 irregularities are not within the geomagnetic low-latitudes. Figure 3's panel D shows Corr-Cluster-4 irregularities. The average profile plot shows that Corr-Cluster-4 irregularities have peak amplitudes of 12 units over latitudes 5N and 20N. The occurrence frequency plot of Corr-Cluster-4 shows that Corr-Cluster-4 profiles are mostly found over the American sector from January till June. The green and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-4 irregularities are partially within the geomagnetic low-latitudes.

Figure 3's panel E shows Corr-Cluster-5 irregularities. The average profile plot shows that Corr-Cluster-5 irregularities have peak amplitudes of around 8 units over latitudes 30S and 20S. The occurrence frequency plot of Corr-Cluster-5 shows that Corr-Cluster-5 profiles are mostly found over the Asian sector from April to August as well as over the American sector from August to April. The green and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-5 irregularities are very far from the geomagnetic low-latitudes.

Figure 3's panel F shows Corr-Cluster-6 irregularities. The average profile plot shows that Corr-Cluster-6 irregularities have peak amplitudes of 9 units over latitudes 30S and 25S. The occurrence frequency plot of Corr-Cluster-6 shows that Corr-Cluster-6 profiles are also mostly found over the Asian sector from April to August as well as over the American sector from August to April. The green and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-6 irregularities are very far from the geomagnetic low-latitudes.

Figure 3's panel G shows Corr-Cluster-7 irregularities. The average profile plot shows that Corr-Cluster-7 irregularities have peak amplitudes of 6 units over latitudes 10N and 25N. The occurrence frequency plot of Corr-Cluster-7 shows that Corr-Cluster-7 profiles are mostly found over the American sector between April and August. The green and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-7 irregularities are very far from thegeomagnetic low-latitudes.

Figure 3's panel H shows Corr-Cluster-8 irregularities. The average profile plot shows that Corr-Cluster-8 irregularities have peak amplitudes of 4 units over latitudes 20N and 30N. The occurrence frequency plot of Corr-Cluster-8 shows that Corr-Cluster-8 profiles are mostly found over the American sector between April and August. The green and red lines indicate that the geographic latitude of the peak amplitudes in Corr-Cluster-8 irregularities are very far from the geomagnetic low-latitudes.

Unlike the results of the Square Euclidean cluster analysis, only one cluster, Corr-Cluster-1 has occurrence frequency plots showing ion density irregularities frequently occurring in months and longitudinal sectors when and where EPBs frequently happen. The latitudinal location of Corr-Cluster-1 irregularities' peak amplitude is also within the geomagnetic low-latitudes. The other clusters' occurrence frequency plots indicate that most of the irregularities in the clusters don't occur in months and longitudinal sectors when and where EPBs frequently happen.

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4. Discussion and Conclusions

This work explores the results of applying Square Euclidean and Correlation k-means cluster analysis on ion density irregularity profiles observed by the Advanced Ionospheric Probe (AIP) onboard the Formosat-5 (F-5) satellite from November 2017 to November 2020. To isolate the ion density irregularities, the profiles were first filtered to give amplitude profiles whose regions of highest amplitudes coincide with the strongest irregularities. These were then subject to the cluster analyses. Of the 8 clusters, the Square Euclidean k-means cluster analysis was able to classify the irregularities into 4 clusters that exhibited characteristics consistent with irregularities

due to EPBs. 2 clusters namely Sq-Cluster-1 and Sq-Cluster-2 mostly comprised of ion density 336 irregularities occurring over the geographic equator and during equinox seasons. They also 337 occurred over Eastern American sector where the geographic equator can be found within the 338 339 geomagnetic low-latitudes. The only difference between the 2 clusters is in the amplitudes. Another cluster labelled Sq-Cluster-3 mostly comprised of ion density irregularities occurring over 340 341 the southern geographic low-latitudes and during equinox seasons. They also occurred over the 342 Western American sector where the southern geographic low-latitudes can be found within the 343 geomagnetic low-latitudes. Finally, Sq-Cluster-4 mostly comprised of ion density irregularities occurring over the northern low-latitudes and during equinox seasons. They also occurred over 344 Atlantic sectors where the northern geographic low-latitudes can be found within the geomagnetic 345 low-latitudes. 346

Of the 8 clusters, the Correlation k-means cluster analysis was only able to yield one cluster that had characteristics consistent with EPBs. This was Corr-Cluster-1. Similar to both Sq-Cluster-1 and Sq-Cluster-2, Corr-Cluster-1 mostly comprised of ion density irregularities occurring over the equator and during equinox seasons. The Correlation k-means cluster analysis was also able to yield clusters that have, for example, peaks over the northern and southern low-latitudes. However, the longitudinal location of the peaks isn't within the geomagnetic low-latitudes.

Previous studies grouped ion density profiles in terms of hyper-parameters such as ion density dip magnitude, depletion edge criteria or length of depletion (Huang et al, 2001; Burke et al, 2004; Gentile et al, 2006). However, the chosen values of these parameters are based on looking at just a few ion irregularity profiles. With a cluster analysis, the algorithm looks at all of the ion irregularity profiles before grouping them. This is a more data-driven approach and is thus more advantageous. These results suggest that a cluster analysis preferably a Square Euclidean cluster analysis can be used to find and classify ion density irregularities consistent with EPBs. This work also shows that the F5/AIP is able to make multi-year observations of ion density irregularities at ~710 km altitude and at ~2230 pm local-time that are consistent with irregularities due to EPBs.

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429

430 **Figure captions**

Figure 1: A) Formosat-5 AIP orbital profiles for one day. B) Formosat-5 AIP orbital profiles for
two days. C) Unfiltered and filtered sample AIP ion profile as well as the amplitude calculated
from the filtered profile as a function of geographic latitude. D) Total number of profiles for each
longitude-month bin accumulated between November 2017 and November 2020.

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Figure 2: Each panel corresponds to a cluster (e.g. cluster-1 is in panel A, cluster-2 is in panel B, 436 etc). The left plot of each panel shows the average of all profiles under the given cluster as a 437 function of geographic latitude. The errorbars are the standard deviation of the profiles. Note that 438 the y-axis for the left plots differ per panel. The right plot of each panel shows the number of 439 440 profiles found in each month and longitude bin. The red line in each panel's left plot is situated over the average geographic latitude of magnetic inclination angle 0° for the given cluster. The 441 green line to the left (right) of the red line in each left plot is situated over the average geographic 442 latitude of magnetic inclination angle -15° (15°) for the given cluster. See text for more details. 443

Figure 3: Same as figure 2 but for the Correlation cluster analysis.