

Determining the Spectral Content of MOSES Images

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ABSTRACT

The Multi-Order Solar Extreme Ultraviolet Spectrograph (MOSES) sounding rocket was launched from White Sands Missile Range on February 8th, 2006 to capture images of the sun in the He II 303.8 Å emission line. MOSES is a slitless spectrograph that forms images in multiple spectral orders simultaneously using a concave diffraction grating in an effort to measure line profiles over a wide field of view from a single exposure. Early work on MOSES data showed evidence of solar features composed of neither He II 303.8 Å or the nearby Si XI 303.3 Å spectral lines. We have built a forward model that uses co-temporal EIT images and the Chianti atomic database to fit synthetic images with known spectra to the MOSES data in order to quantify this additional spectral content. Our fit reveals a host of dim lines that alone are insignificant but combined contribute a comparable intensity to MOSES images as Si XI 303.3 Å. In total, lines other than He II 303.8 Å and Si XI 303.3 Å contribute approximately 10 percent of the total intensity in the MOSES zero order image. This additional content, if not properly accounted for, could significantly impact the analysis of MOSES and similar slitless spectrograph data, especially those using a zero order image.

1. INTRODUCTION

An essential tool used to study remote solar phenomena is the spectrograph. By more closely examining the wavelength of light emitted from the sun we gain access to a wealth of information (temperature, density, velocity, etc.) needed to more completely describe the solar plasma and properly inform our models. Slit spectrographs take a very narrow strip of the sun and disperse the light emitted using a diffraction grating, providing high spectral resolution over a very small spatial field of view. Spectrally resolved images are then built by rastering the slit over a region of interest, requiring a full exposure at every position, which can take on the order of hours to cover a large portion of the solar disk.

An alternative approach is to remove the slit, a *slitless* spectrograph. Slitless spectrographs disperse an image over a larger field of view. The data from these instruments, sometimes called overlapograms, have spatial and spectral information overlapped in the dispersive direction. Slitless spectrographs often image in multiple spectral orders, each with a different blend of spatial and spectral information, that can be combined and “inverted” to form a spectrally resolved image of the sun at every exposure. By doing so these instruments trade spatial and spectral resolution for much higher cadence.

Slitless spectrographs have been used infrequently for decades to image the sun and have recently regained popularity. Only two satellite missions have routinely captured solar slitless spectrograph data: The S082A instrument on *SkyLab* (Tousey et al. 1973) and the Res-K instrument of the Russian

KRONOS-I mission (Zhitnik et al. 1998). These two instruments have informed the recent development and proposal of another full disk slitless spectrograph, the COronal Spectroscopic Imager in the EUV (COSIE: Winebarger et al. 2019; Golub et al. 2020). The currently-operating Extreme-ultraviolet Imaging Spectrograph (EIS: Culhane et al. 2007) on the *Hinode* mission (Kosugi et al. 2007) includes 40'' and 266'' slots that can produce overlappogram that has been used in a limited number quantitative studies (Harra et al. 2017, 2020). In addition to these satellite missions several sounding rocket based slitless spectrographs have recently imaged the sun including The Multi-Order Solar Extreme ultraviolet Spectrograph (MOSES: Fox 2011), the Extreme ultraviolet Snapshot Imaging Spectrograph (ESIS: Smart et al. 2021; Parker et al. 2021), and the recently launched MARshall Grating Incident X-ray Spectrometer (MAGIXS). JDP: Not sure how to cite this, will talk to Amy

MOSES was launched for the first time from White Sands Missile Range on February 8th, 2006. The primary science goal of MOSES was to measure line profiles in He II 303.8 Å over a wide field of view ($\approx 20' \times 10'$), at every exposure. By imaging in multiple orders simultaneously, $m = -1, 0$, and 1 , MOSES captures three different projections through a spatial-spectral cube, $I(x, y, \lambda)$, that can be combined and inverted to return a line profile at every pixel over its large field of view, at every exposure. Various inversion methods have been used to return line profiles thus far, mostly for small explosive events in He II 303.8 Å (Fox 2011; Courier & Kankelborg 2018; Rust & Kankelborg 2019).

Early work with MOSES images also revealed solar features when subtracting different spectral orders, as well as in the residuals while inverting, that could not be attributed to He II 303.8 Å. These features could not be attributed to the most obvious sources of spectral contamination, including the nearby Si XI 303.3 Å, or two bright Iron lines, Fe XV 284 Å and Fe XVI 335 Å. The periodic multi-layer coatings used on the MOSES primary and secondary optics were designed to allow for optimal throughput at 304 Å but to suppress the close Iron lines, Fe XV 284 Å and Fe XVI 335 Å, as much as possible (Owens et al. 2005) making them unlikely source of contamination. Si XI 303.3 Å, which is too close in wavelength to be removed by the MOSES multi-layer coating was always expected to be a contamination source in the data. With a spectral dispersion of $\approx 29 \text{ mÅ}$ per pixel ($\approx 30 \text{ km s}^{-1}$ per pixel), features in Si XI 303.3 Å are dispersed roughly 15.7 pixels from He II 303.8 Å and therefore are easily distinguishable features in He II 303.8 Å. Therefore these unexpected solar features must be from emission at dimmer and less obvious wavelengths in the MOSES passband. In this paper we identify and quantify the sources of spectral contamination in the MOSES images. We also comment on the MOSES design decisions that led to this confusion, and how similar instruments can be adapted to better control spectral contamination.

In Section 2 we present a time averaged set of images from the 2006 MOSES flight, as well as 4 co-temporal and co-aligned images from SOHO EIT. We also discuss the unexpected solar features of unknown spectral content found in the MOSES data. Section 3 describes our methods for identifying and quantifying sources of spectral contamination. This includes using the cross-correlation of MOSES difference images as an indication of significant spectral contamination (Section 3.1 and 3.2) and how we combine data from EIT with synthetic spectra from the Chianti database (Dere et al. 1997; Del Zanna & Young 2020) to generate synthetic MOSES images that can be fit to MOSES images (Section 3.3). Section 4 shows our best fit to the MOSES data, the resulting spectral content of the time averaged MOSES images. We discuss our results and their implications for future missions of this kind in Section 5.

2. DATA

2.1. *MOSES*

The MOSES sounding rocket launched from White Sands Missile Range on February 8th, 2006 at 18:44 UT. It recorded 27 exposures between 18:44:17 and 18:49:13 UT above 160 km in altitude. Exposure times ranged from .25 - 24 seconds with a roughly 6 second readout time in between. An exposure consists of three images, one for each of the three spectral orders $m = -1, 0$, and 1 . For the rest of the paper we will refer to the m th order MOSES image as, I_m . All data was dark subtracted and co-aligned to exposure number 13 prior to our study (Fox 2011).

Of the different exposures, longer exposures are best for observing quiet sun features but are saturated near active regions. To fill in missing saturated regions and increase signal to noise we form a single time averaged image in all three spectral orders. Saturated pixels are masked with NaNs (Not a Number) and treated as missing data. Since MOSES observes through changing amounts of atmosphere throughout its flight we use the median of each image as a synthetic exposure time, rather than the amount of time the shutter is open. Masked data is then summed in time and divided by a total synthetic exposure time for each pixel to form a single time averaged image in each spectral order with no saturated pixels. The top row of Figure 1 shows the time averaged version of I_1 and I_0 .

Identifying solar features in the MOSES data that have undergone spectral dispersion is simple. Subtracting I_0 (that contains no spectral dispersion) from either outboard order (I_{+1} or I_{-1}) eliminates undispersed features. What remains are bipolar features of various spatial scales. A feature in the principle wavelength, He II 303.8 Å, with a nonzero line-of-sight (LOS) velocity is translated along image rows. MOSES has a spectral dispersion of $\approx 30 \text{ km s}^{-1}$ per pixel, leading to a less than ten pixel shift for even the fastest LOS velocities in He II. This results in a bipolar feature with an obvious positive and negative counterpart that are immediately next to one another. Events like these have been studied in detail by several authors (Fox 2011; Courier & Kankelborg 2018; Rust & Kankelborg 2019).

Features from other emission lines in the MOSES passband have shifts larger than 10 pixels and cannot be mistaken as Doppler shifted features in He II 303.8 Å. A feature in Si XI 303.3 Å, the next closest line, would be shifted by 15.7 pixels. Si XI features can be seen on the solar limb where He II has little to no contribution. The best example of this is region 3 boxed in red in Figure 1. Box 1 of Figure 1c has a large, coherent, negative feature dubbed the “wishbone”. The wishbone is very solar in appearance and has no obvious positive counterpart. Close inspection reveals a white smear to its left that is likely a shifted wishbone in the plus order. Unfortunately the positive portion of the wishbone is too blurry to quantify its shift by inspection. A previous study of these features by Rust (2017) used a wavelet transform to isolate large scale features prior to taking the difference. That procedure allowed Rust (2017) to roughly identify a contribution from Mg VIII $\lambda 315$ Å to regions 1 and 2. Rust (2017) also identified a faint copy of the limb, $\approx 200''$ to the right of region 3 that was attributed to Si IX/ Fe VI 296.1 Å. Although the probable sources of contamination were identified, their contributions to the total image intensity were not quantified.

2.2. *EIT*

In support of the 2008 MOSES rocket flight the Extreme ultraviolet Imaging Telescope (EIT: Delaboudinière et al. 1995) on board the Solar and Heliospheric Observatory (SOHO) captured 4 full disk EUV images, one in each of the 171 Å, 195 Å, 284 Å, and 304 Å channels (Figure 2). Prior

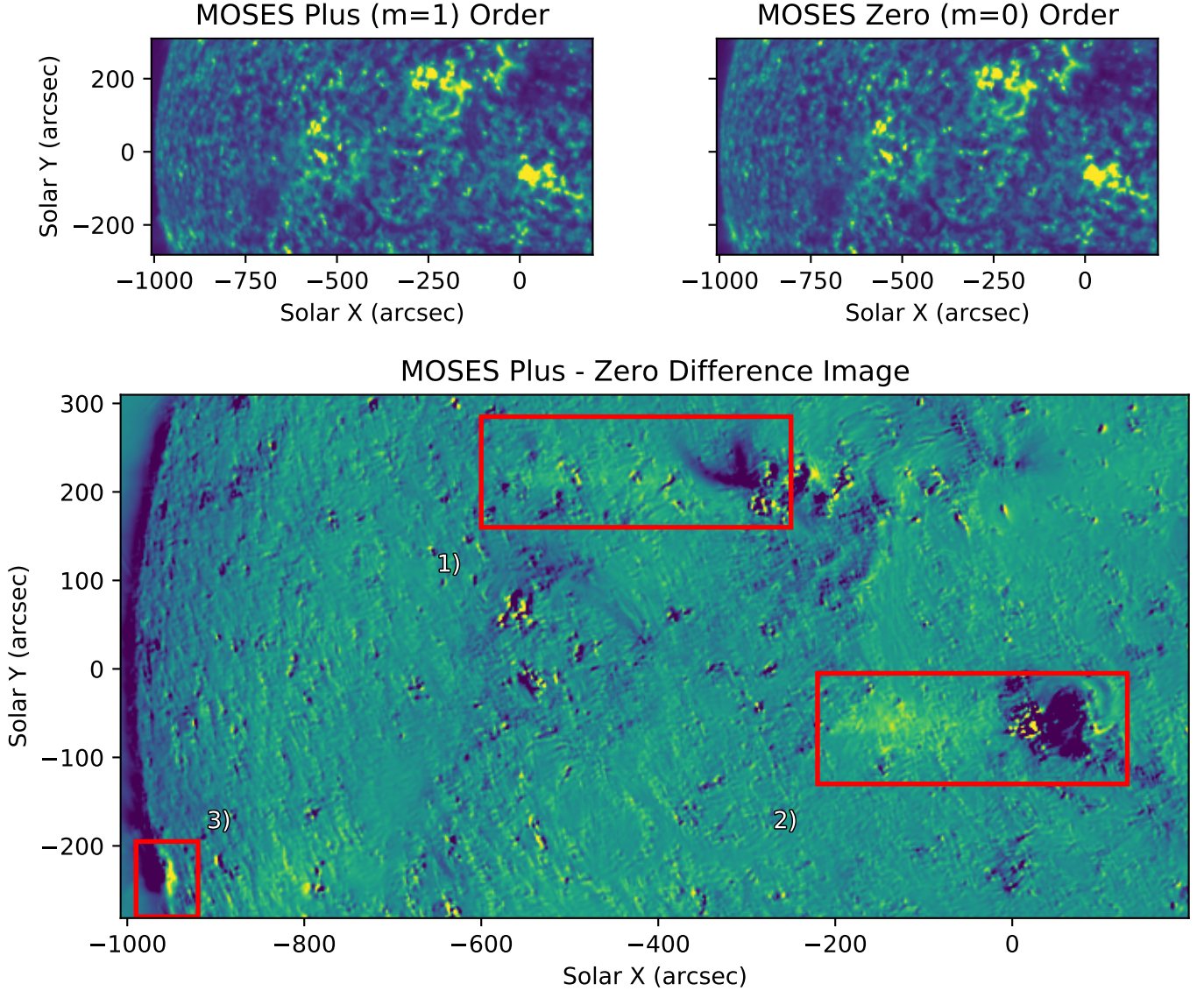


Figure 1. Time averaged images for the $m = 0$ and 1 spectral orders (top row), followed by their difference, $m = 1$ minus $m = 0$. Regions one through three, boxed in red, show regions of high spectral contamination. Dark features from the $m = 0$ order are adjacent to white smears with pixel shifts too large to be Doppler shifts.

to use, each image was first despiked using `iris_prep_despike.pro` with default settings and made level 1 with `eit_prep.pro`. They were then rotated to 2006-02-08 18:47 UT, the timestamp of image 13 of the MOSES observing sequence (Fox 2011), using `drot_map.pro`. EIT and MOSES were co-aligned by applying the affine coordinate transform to EIT 304 Å that maximized the zero-lag cross-correlation between it and the time averaged $m=0$ order image described in the proceeding subsection. The same transform was then applied to the other three EIT channels.

At first glance one can find several similar feature in the MOSES difference image (Figure 1) and the EIT 171, 195, and 284 images (Figure 2). The presence of these features in MOSES data indicates a contribution of coronal spectral lines to an otherwise transition region image. While similarities between MOSES difference images and EIT images indicate the contribution of hotter lines to MOSES

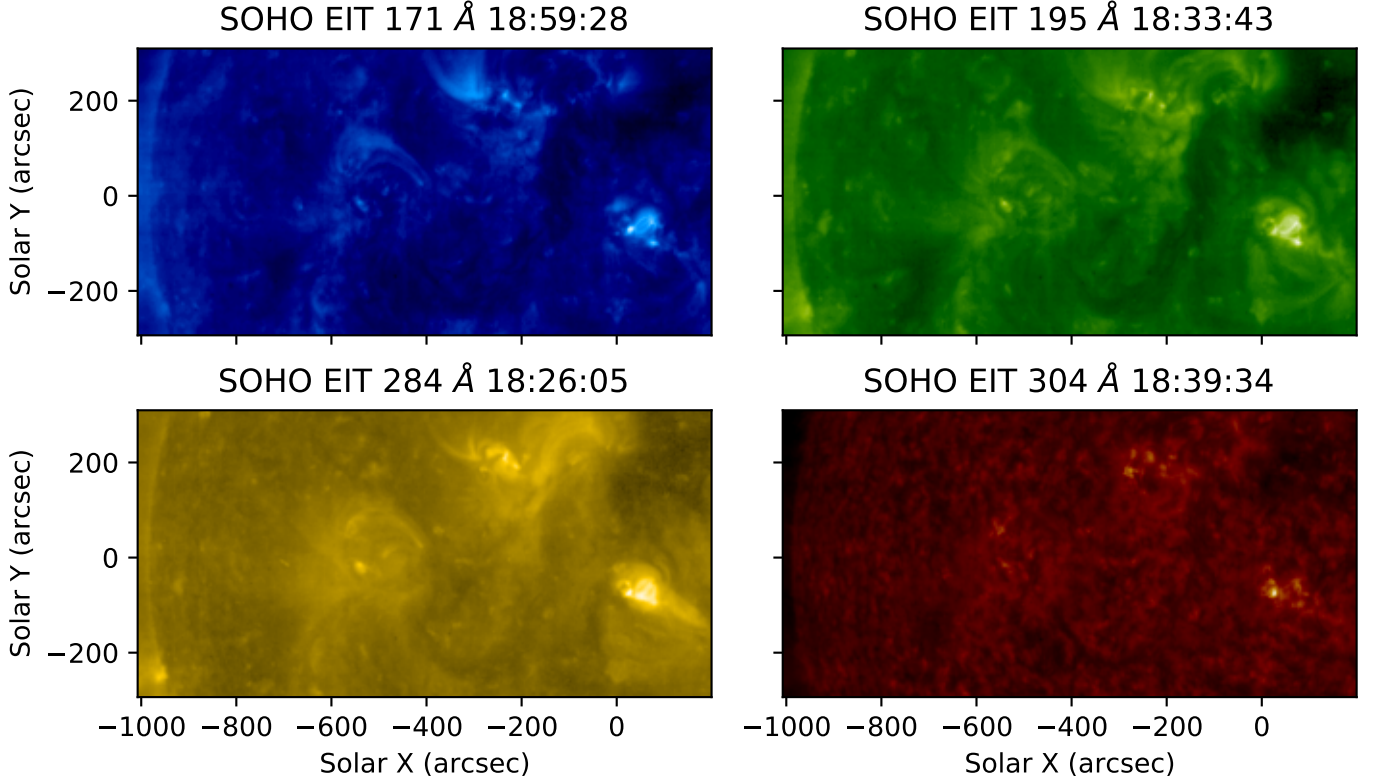


Figure 2. SOHO EIT Images taken closest the MOSES Launch on February 8th, 2006. Each image was rotated to 2006-02-08 18:47 UT to match the middle MOSES exposure time. After rotation, the EIT 304 channel was linearly co-aligned to the MOSES zero order, and then the same transformation was applied to every other channel. Each image has been fourth root scaled.

He II data, they do not tell how much, and by which lines. In Section 3 we set out to quantify the intensity contributed by each source of spectral contamination in the MOSES image.

3. METHODS

3.1. Difference Image Cross-Correlation

To help identify subtle pattern repetition in the MOSES difference images we cross-correlated them along the dispersion direction, or image rows. Performing a cross-correlation on the difference images requires justification. An obvious first choice would have been to simply cross-correlate I_0 with either outboard order. Unfortunately the correlation function is dominated by the autocorrelation of the He II signal as seen in Figure 3d and 3e. These example cross-correlations show two main peaks in correlation that are due to the auto-correlation of bright He II features, and not spectral contamination. This would also be the case when cross-correlating I_{+1} and I_{-1} . By taking the difference we remove stationary He II objects (Figure 4d and 4e) and, in turn, their autocorrelation from the cross-correlation function (Figure 4f), leaving only peaks in correlation from features not of He II 303.8 Å.

The cross-correlation of two difference images along their rows is defined to be,

$$I_{+1} - I_0 \otimes I_{-1} - I_0 = \mathcal{F}_x^{-1} \{ \mathcal{F}_x (I_{+1} - I_0) * \mathcal{F}_x (I_{-1} - I_0) \}, \quad (1)$$

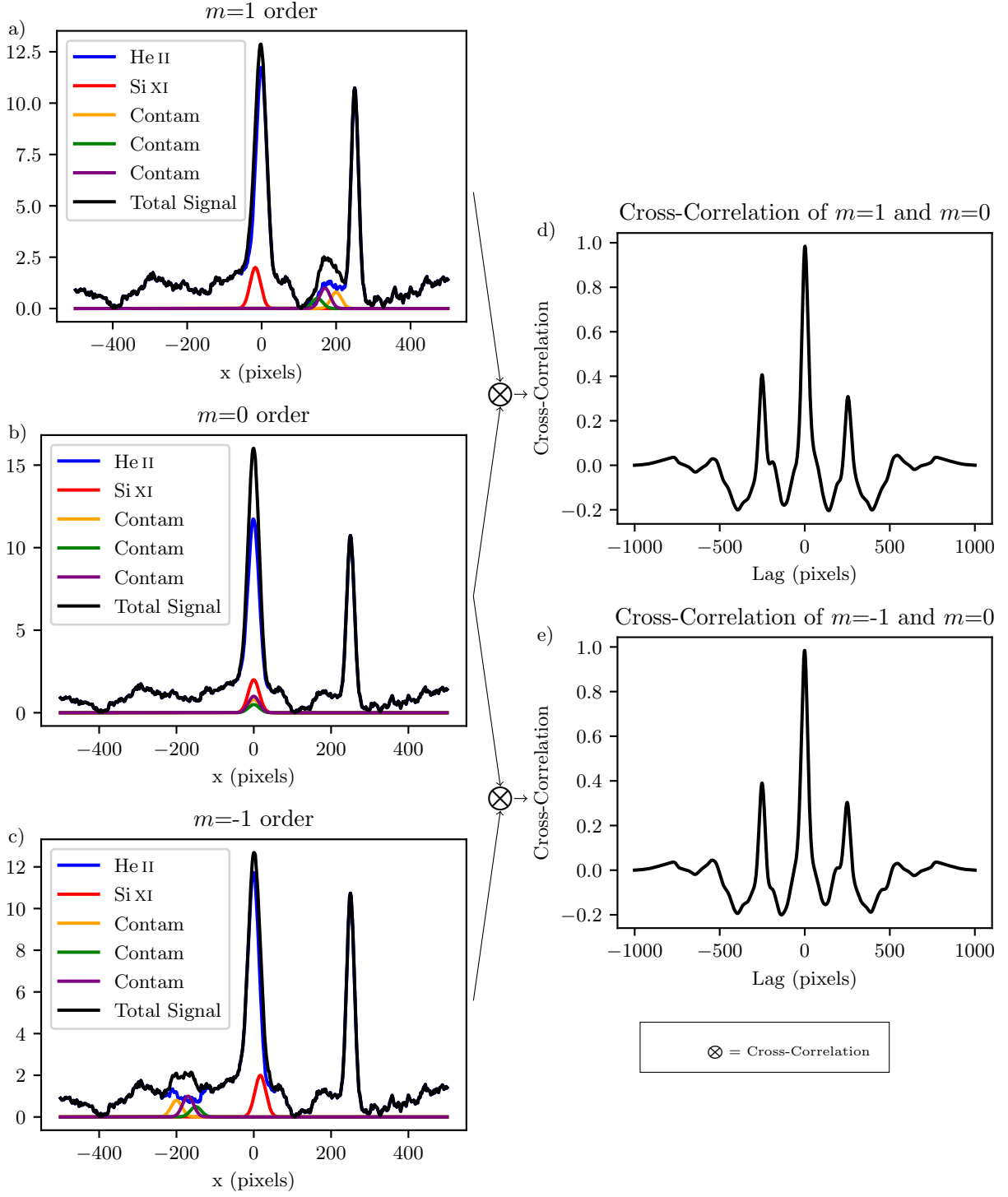


Figure 3. Example 1D MOSES row in the $m = 1$ (a), 0 (b), and -1 (c) spectral orders followed by the cross correlation of $m=1$ and 0 (d) and the cross-correlation of $m = -1$ and 0 (e). The cross-correlation of any two MOSES orders is dominated by the autocorrelation of stationary He II features and background making them less useful in identifying spectral content. The strong peaks in correlation at ± 250 pixels are the result of the two bright He II features at 0 and 250 overlapping.

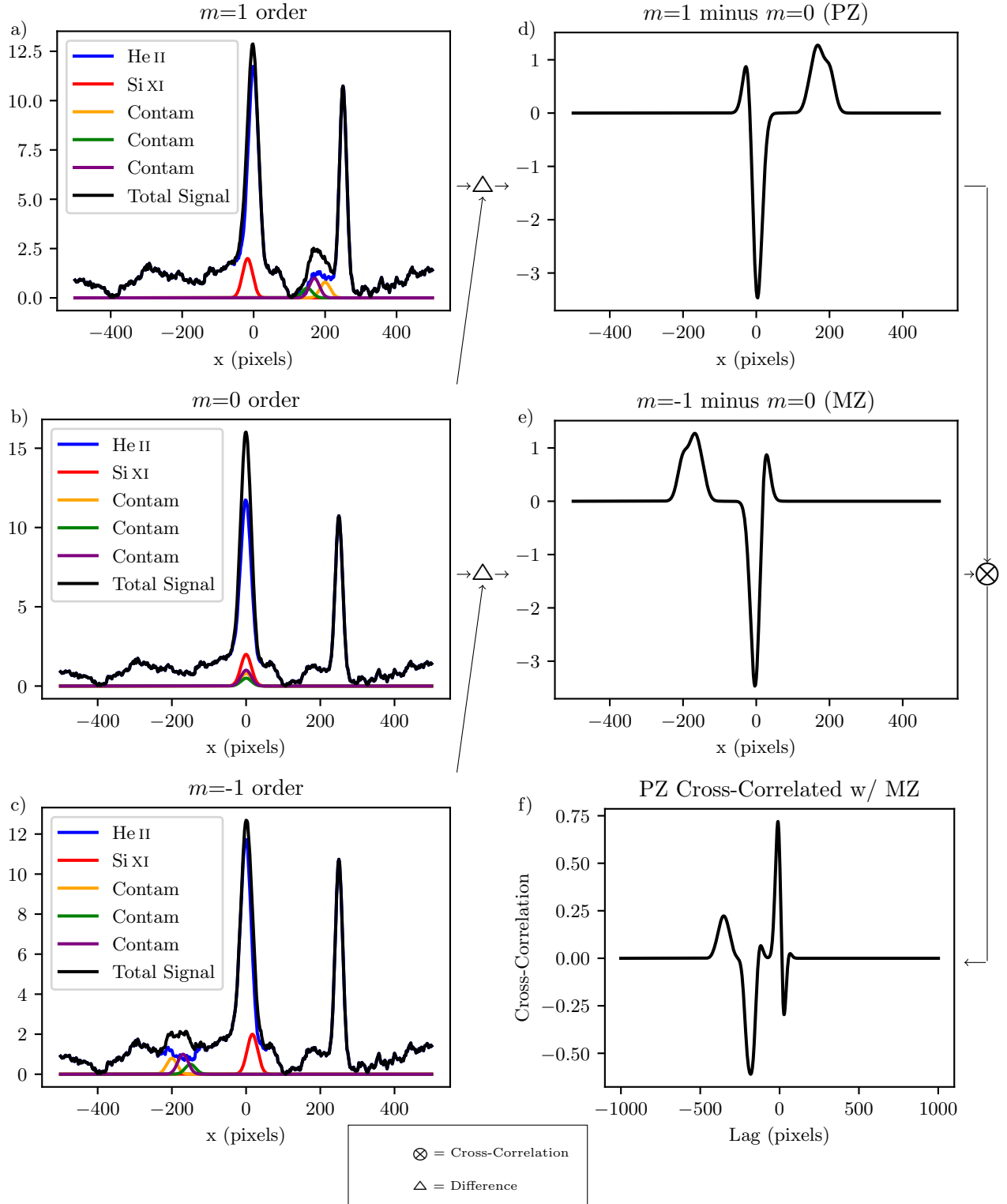


Figure 4. Example MOSES rows in $m = 1$ (a), 0 (b), and -1 (c) spectral orders. Two difference images, $m=1$ minus $m=0$ (d) and $m=-1$ minus $m=0$ (e) are cross-correlated (f). Taking a difference removes stationary He II features so that peaks in cross-correlation are indicative of extra spectral content. Even in this simple example one cannot read off the spectral content of the MOSES images, demonstrating the need for a forward model.

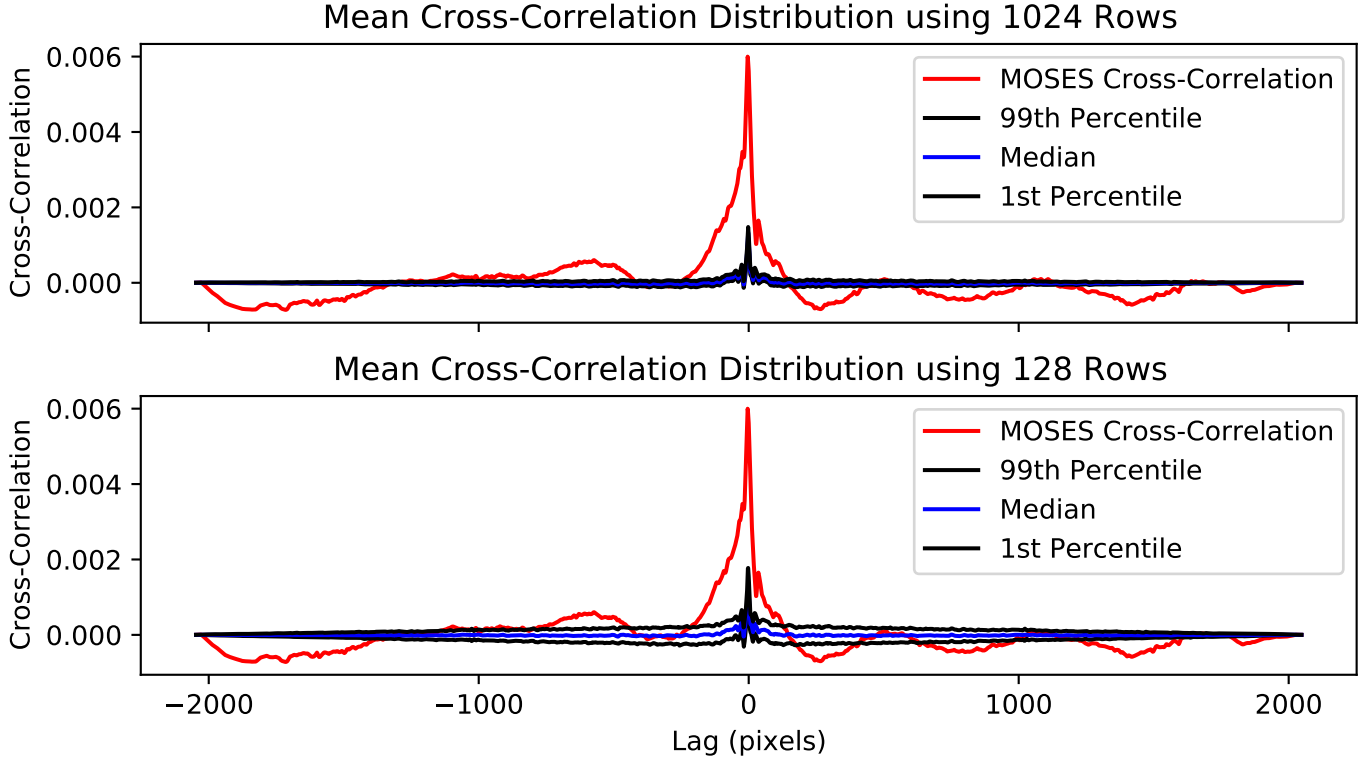


Figure 5. The mean cross-correlation function of MOSES difference image rows (red) is compared to a distribution of mean cross-correlations from a selection of randomly generated test row differences (Section 3.2). The top panel shows the distribution after taking the mean of 1024 (the number of MOSES rows) randomly chosen test row differences 10,000 times. The bottom panel challenges that all 1024 rows of MOSES are unique and shows the distribution for only 128 randomly chosen difference rows, also 10,000 times. In either case the mean cross-correlation values from the MOSES data exceed the 98 percent confidence interval and are therefore deemed significant and indicative of spectral content in the MOSES data.

where \mathcal{F}_x is the Fast Fourier Transform (FFT) operator applied along image rows. Prior to subtracting each image is normalized by its median. Also, image rows were padded with zeros, prior to applying the FFT to minimize edge effects. This yields a one dimensional cross-correlation function for each row of the MOSES difference images. Since we are concerned mostly with bulk spectral content we then take the mean of all 1024 cross-correlation functions, one for each row, to get our final correlation curve plotted in red in Figure 5.

The mean cross-correlation curve for the MOSES difference image has a few notable features. There are larger peaks in correlation at approximately -1800, -500, 250, 800, and 1500 pixel lags. The largest peak in correlation is centered about zero lag and is quite broad. While these features are identifiable, the curve is complicated enough that quantifying peaks in correlation visually is difficult and the significance of any given peak is questionable. We therefore move to test the null hypothesis that none of these features are the due to spectral contamination but are instead the result of random fluctuations in the data by cross-correlating random data generated to match the MOSES image rows.

3.2. Significance Testing

The mean cross-correlation function of the two MOSES difference images, red in Figure 5, has several interesting peaks in correlation. As can be seen in Figure 4e, these peaks can be indicative of extra spectral content. Actual MOSES images contain information not represented in the Figure 4 cartoon. Each MOSES order has a slightly different point spread function and the He II features are not all stationary. This leads to many extra small scale features in the difference images across the field of view. It is also unclear what magnitude of cross-correlation we should expect from extra spectral content. We therefore move to reject the null hypothesis that the features in the MOSES difference image cross-correlation curve are the result of random correlations between each difference image and not indicative of extra spectral content.

To attempt to reject the null hypothesis we cross-correlated randomly generated test solar data that match the MOSES image rows in length and have similar power spectral and autocorrelation distributions. Our test data set was generated from the MOSES image columns because they contain the same solar features as the image rows, but contain practically no spectral dispersion. By interpolating and shuffling the elements of the MOSES columns in Fourier space we can generate a large number of test arrays for significance testing.

MOSES images contain 2048 columns. These columns have the same spatial features as MOSES image rows but with none of the spectral information. Therefore, cross-correlating MOSES image columns will show the expected magnitude of cross-correlation from non dispersed solar features. Despite this the MOSES image columns are insufficient for significance testing in a couple ways. First, there is an insufficient sample size. With features in the MOSES images ranging from ≈ 4 -100 pixels in size we have at most 512 unique columns for significance testing. Second, MOSES columns are half as long as the rows, preventing us from measuring the significance of correlation past 1024 pixel lag. Therefore we needed to generate a test data set for significance testing.

Using the MOSES image columns as our basis we generated N random arrays that are 2048 pixels long and match MOSES columns in both power spectral and autocorrelation distribution. First the columns in each time averaged MOSES image, I_{mxy} , are windowed with a Hanning window, w_y , and Fourier transformed along the column dimension, y . The windowed Fourier transformed array is defined to be

$$\tilde{I}_{mxk_y} = \mathcal{F}_y [w_y I_{xy}] \quad , \quad (2)$$

and the Hanning window is defined as,

$$w_y = \alpha - (1 - \alpha) \cos \left(\frac{2\pi y}{N} \right) \quad , \quad (3)$$

where N is the number of elements in y and $\alpha = .5$. The spatial dimensions x and y represent MOSES image row and column positions respectively with $x = 0, 1, \dots, 2047$ and $y = 0, 1, \dots, 1023$. Image columns were windowed prior to the Fourier transform so that the test data could be compared to the MOSES data with less influence from the discontinuity at the image edge.

From the set of all the Fourier transformed data columns, we will generate a test *row* for each order, I'_{mx} , by populating its Fourier transform, \tilde{I}'_{mk} . Each new array is formed by picking an element randomly from the distribution of Fourier transformed columns. The transformation outlined in Equation 2 and 3 gives a distribution of 511 spatial frequency bins and one DC bin that each have 1024 elements (one from each column) for each order. Once assembled, each array is further scrambled

by giving each value of k , aside from the DC term, a random phase shift, $e^{i\phi}$. By this method the k^{th} element in the n^{th} test row, \tilde{I}'_{mkn} , is

$$\tilde{I}'_{mkn} = \tilde{I}_{m\sigma k_y} e^{i\Phi}, \quad (4)$$

where σ is a random integer between 0 and 2047 and Φ is a random phase between 0 and 2π , both uniformly distributed and chosen n times.

In order to create an array that is 2048 elements long from one that is 1024 elements long we require twice as many values of k . We solve this problem by double picking values from the distribution of k_y for each wave number, k . The values of k_y used in Equation 4 are,

$$k_y = \text{floor}(k/2), \quad (5)$$

where $k = 1, 2, \dots, 1022$ and **floor** rounds down to the nearest integer. Since our data is purely real we can fill in the remaining Fourier components, negative frequencies, with the complex conjugate of the corresponding positive frequency component. The final n^{th} test MOSES row of spectral order m is found through an inverse FFT,

$$I'_{mxn} = \mathcal{F}_x^{-1} \left[\tilde{I}'_{mkn} \right]. \quad (6)$$

To verify that our n test arrays match the MOSES columns we take a randomly selected 1024 element long section of each test array, as well as the MOSES columns, and compare their power spectral and autocorrelation distributions. Figures 6 and 7 show three percentiles of both power spectral and autocorrelation distributions (10th, 50th, and 90th) for $n = 10,000$ test arrays. Figure 6 shows great agreement between test data and the MOSES columns in power spectral distribution. Figure 7 shows good agreement between the test data and MOSES columns at the median and only marginal agreement in the wings of the distribution. Despite that the test rows most always has a higher autocorrelation length than the MOSES columns and therefore acts as a worse case scenario during significance testing.

Now that we have created a set of n test rows for each order m we can begin to cross-correlate them and compare the results to the mean MOSES difference image cross-correlation curve. Every test row in each order, I'_{mxn} , is normalized by its median, and then subtracted and cross-correlated according to Equation 1. This yields a set of $n = 10,000$ test difference row cross-correlation functions. Since we are interested in the average MOSES difference row cross-correlation we choose a number of test row cross-correlations and average them together before comparison. The top panel of Figure 5 shows the result of picking 1024 random test cross-correlations (the same number as a MOSES image) and taking their average. We do this 10,000 times and form a distribution of mean cross-correlation values at each pixel lag. The 1st and 99th percentile of the distribution are plotted in blue and the median in black. From the top panel of Figure 5 it is clear that a mean cross-correlation much greater or less than zero is highly unlikely in our test rows with no spectral content. The exception to this is at zero lag when a positive peak in correlation well above zero is typical. Despite this exception the peak in mean cross-correlation from the MOSES data is much greater, and broader, than that of the test data.

The comparison in the top panel of Figure 5 assumes that all 1024 difference rows from the MOSES data are unique. Since solar features in the MOSES images exceed a single pixel in size this is unlikely. The bottom panel of Figure 5 is made in the same way as the top panel but uses the mean of only 128

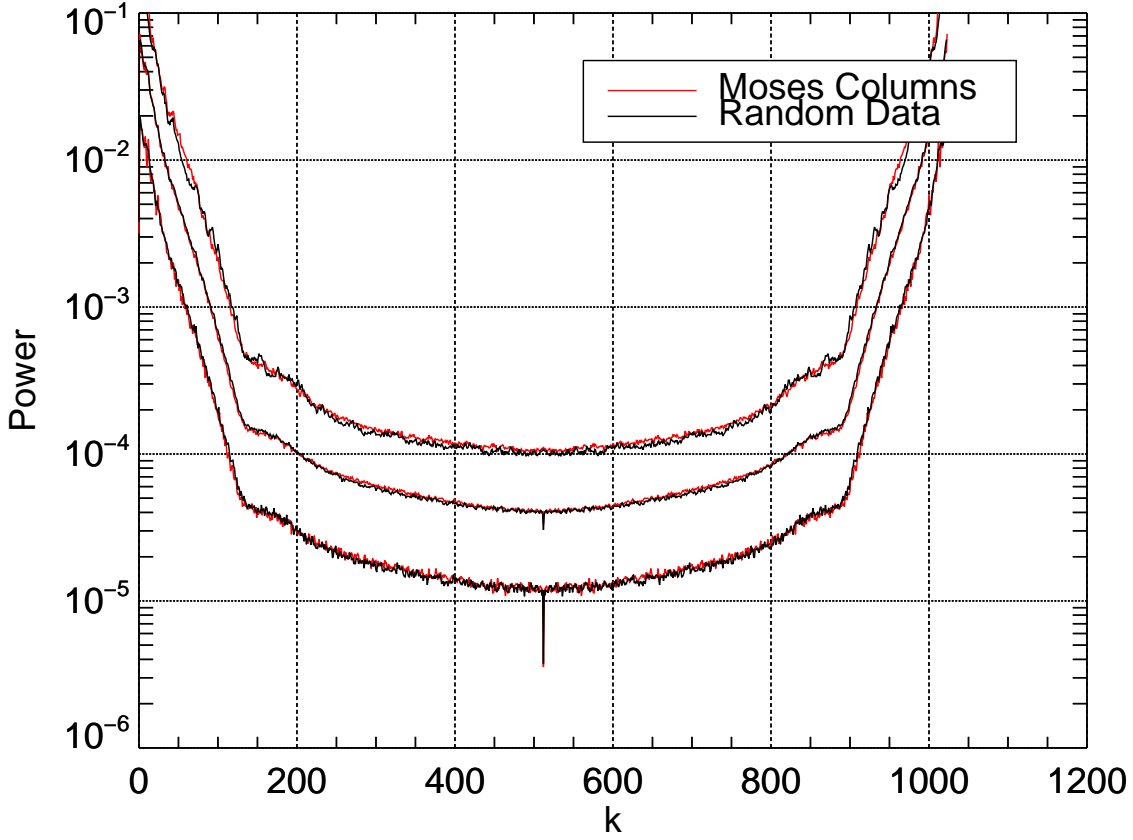


Figure 6. The 10th, 50th, and 90th percentile of the power spectral distribution for each value of k is plotted for both the $m = 0$ MOSES columns and test data.

rows when forming the distribution. Even when we assume that MOSES only 128 unique rows we can see the same peaks in the MOSES mean cross-correlation fall outside the 98 percent confidence interval.

From this study we find that the shape and magnitude of the peaks in the mean cross-correlation of the MOSES difference images are significant and indicative of spectral content. In the following section we use a forward model of the MOSES instrument to form synthetic images with a known spectral content and fit them to the MOSES data in order to identify and quantify the sources of spectral contamination.

3.3. Forward Model

The mean cross-correlation of MOSES difference image rows is shown in Figure 5. In the proceeding section several peaks in correlation were deemed to be significant and indicative of spectral contamination in the MOSES data. These peaks have irregular, broad profiles and are both positive and negative. In order to interpret how these peaks in correlation relate to spectral content of the MOSES

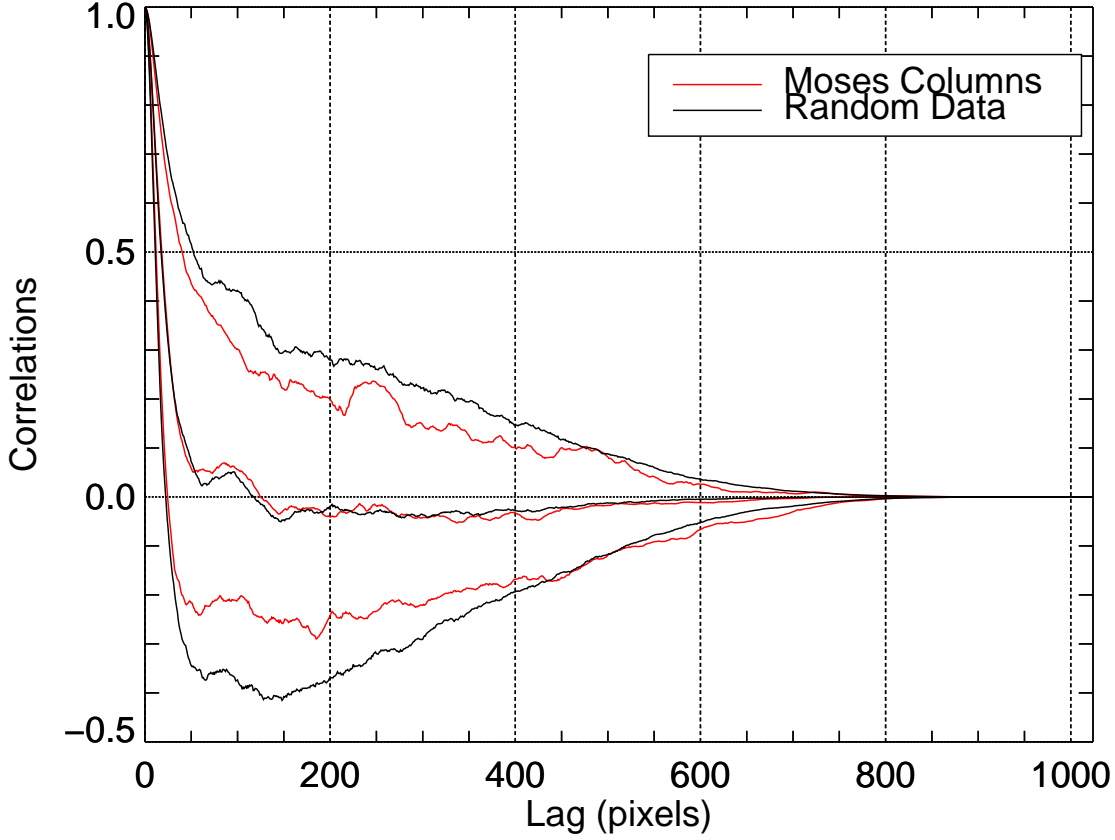


Figure 7. The 10th, 50th, and 90th percentile of the auto-correlation distribution for each lag is plotted for both the $m = 0$ MOSES columns and test data.

images we developed a forward model that produces synthetic MOSES difference images with known spectral content using the 4 co-temporal EIT images and Chianti. These synthetic images can be cross-correlated in the same way and compared to the mean MOSES cross-correlation function. Synthetic MOSES images for each order m , I'_{mxy} , were created using the following procedure:

1. Form a DEM map, a DEM for every EIT pixel, from 4 co-temporal and co-aligned EIT images,

$$\text{DEM}_{xyT} \leftarrow \text{EIT}_{i xy} \quad \text{where } i = 0, 1, 2, 3. \quad (7)$$

2. Normalize the DEM map by its average DEM over the MOSES field of view.

$$\text{DEM}_{xyT} \leftarrow \text{DEM}_{xyT} - \frac{\sum_{xy} \text{DEM}_{xyT}}{N}, \quad (8)$$

where N is the total number of MOSES pixels.

3. Calculate the contribution function, $G_{\lambda T}$, over the range $\log T = 4-6.5$ K using the Chianti routine `ch_synthetic.pro`, for each spectral line between 280 Å and 340 Å using a flat emission measure.
4. Shift the DEM map along image rows according to the MOSES spectral dispersion. Sum the dispersed map multiplied by the contribution function and MOSES optical throughput, $\alpha_{m\lambda}$, along the wavelength axis.

$$I'_{mxyT} = \sum_{\lambda} \text{DEM}_{(x-m\frac{\lambda-303.78}{\delta})yT} \alpha_{m\lambda} G_{\lambda T}. \quad (9)$$

5. Multiply by desired DEM and integrate in temperature to form a synthetic MOSES image in each order.

$$I'_{mxy} = \sum_T I'_{mxyT} \text{DEM}_T dT \quad (10)$$

The DEM map, calculated using `eit_dem_tool.pro` provided in Solar Soft, combines the four prepped EIT images described in Section 2.2 and gives a DEM for each pixel across the full EIT field of view. We normalize the DEM map by its average DEM over the MOSES field of view because we are more concerned with the spatial distribution of intensity, at a given temperature, than the absolute DEM predicted by the code. Also, starting with a normalized DEM map allows us to use a each temperature bin in the DEM as the free parameters in our fit.

In order to create a synthetic MOSES image at each spectral order we must combine the EIT DEM map with spectral information. Using the Chianti database we generate a synthetic line list using a constant log pressure of 15 K cm^{-3} , Feldman coronal abundance (Feldman 1992), and a flat log DEM, $\log \text{DEM}(T) = 1 \text{ cm}^{-5} \text{ K}^{-1}$. Leaving the line list unsummed in temperature gives the intensity as a function of temperature for each spectral line in the passband, $G_{\lambda T}$. Combining $G_{\lambda T}$ with the EIT DEM map and the wavelength dependent MOSES throughput, $\alpha_{m\lambda}$, yields an image for every spectral line at every temperature, $I'_{xy\lambda T}$. The MOSES throughput is defined by the periodic metallic multi-layer coatings on the MOSES primary and secondary optics, as well as the thin film aluminum filters just before the detector.

The MOSES grating has a spectral dispersion of $\delta = \approx 29 \text{ mÅ}$ per pixel. Therefore in order to produce a set of images for each of the three MOSES spectral orders, $m = -1, 0, 1$, line images, $I'_{xy\lambda T}$, must be shifted along x. Each image is shifted by n pixels where $n = m\frac{\lambda-303.78}{\delta}$. This results in no shift for the central, $m = 0$ order, and an equal and opposite shift in the outboard orders, $m = -1$ and 1. When shifting line images to form the outboard orders we allow features out side the field of view defined by the MOSES zero order to shift into the field of view, which is possible since each EIT image is full disk. This mimics the behavior of the actual MOSES optical system since MOSES does not have a stop to strictly enforce its FOV, therefore making it is possible for highly shifted features to exist in the outboard orders that aren't captured by the ESIS zero order and vice versa. After the line image at each temperature is shifted we sum each order in wavelength to form an image in each order at every temperature bin, I'_{mxyT} .

Final synthetic MOSES images are formed at each order by multiplying the temperature dependent cube by a DEM of choice and integrating in temperature. This gives 3 synthetic MOSES images from which we can form difference images and cross correlate (Equation 1). Using each temperature bin

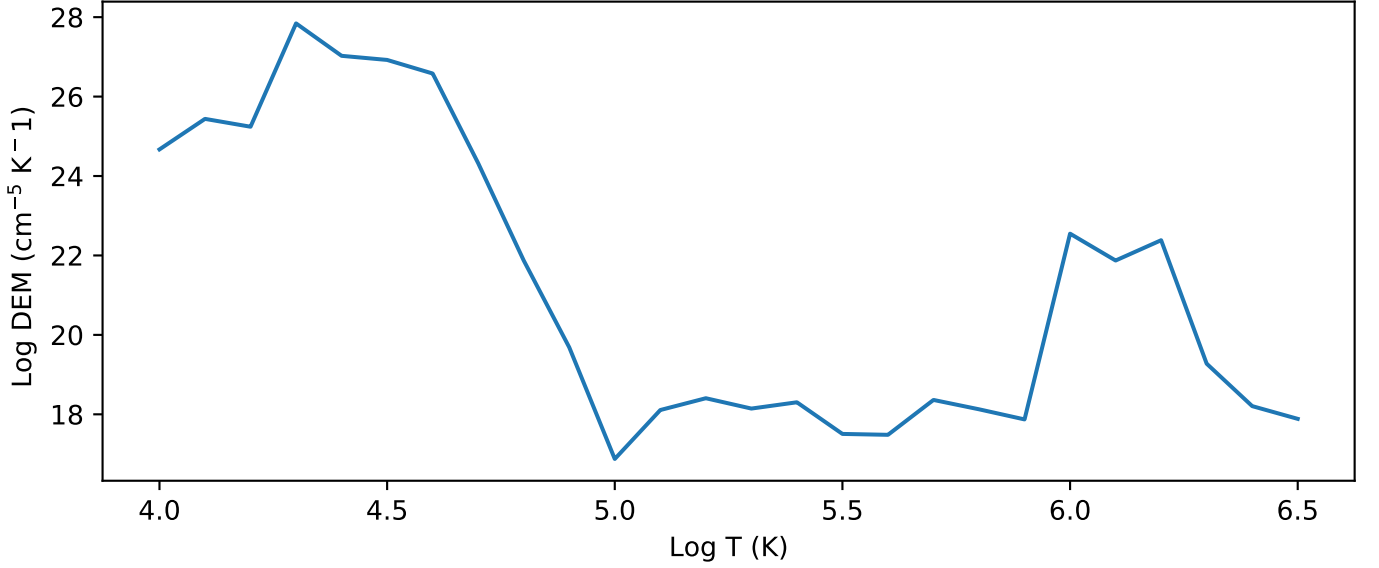


Figure 8. Best fit DEM used to generate the final synthetic MOSES images.

of the DEM as a free parameter, we seek to minimize the least squares difference between synthetic and MOSES mean cross-correlation function, yielding closely matching synthetic MOSES data with known spectral content.

When comparing our synthetic data to MOSES we slightly blur each MOSES image prior to taking the difference and cross-correlating using a normalized Gaussian kernel with a standard deviation of four pixels. This ensures small scale features in the MOSES images (that aren’t indicative of spectral contamination) don’t impact the fit. During testing it was found that the width of the kernel used had little impact on the final results (a few percent difference in final spectral content), therefore a standard deviation of 4 pixels was chosen based on a factor of ≈ 4 difference in resolution between EIT and MOSES, and visual comparison of small scale features in the resulting difference images.

In the following section we show the DEM of best fit to the MOSES data, as well as the resulting synthetic difference images and total spectral content.

4. RESULTS

Using the forward model described in the previous section we have generated synthetic MOSES images with a known spectral content that best fit the MOSES data. We achieved a best fit by varying the emission measure at each temperature prior to integrating, and then comparing the cross-correlation of the synthetic difference images to that of the MOSES difference images. The DEM of best fit is shown in Figure 8.

We find good agreement between our synthetic data of best fit and the MOSES data. Figure 9 compares the mean cross-correlation of the difference image rows from the slightly blurred MOSES data, to same cross-correlation of the synthetic data. Comparing the two curves shows notable agreement in a few places including around zero lag and all positive lags. Negative lags match less closely, although the shape between zero and approximately -600 pixels is similar.

A synthetic difference image of best fit (I'_{+1} minus I'_0) is shown along with the corresponding MOSES difference image in Figure 10. In the synthetic difference image each feature identified in Figure 1 has been successfully reproduced. The dark imprint of the “wishbone” (region 1) and nearby

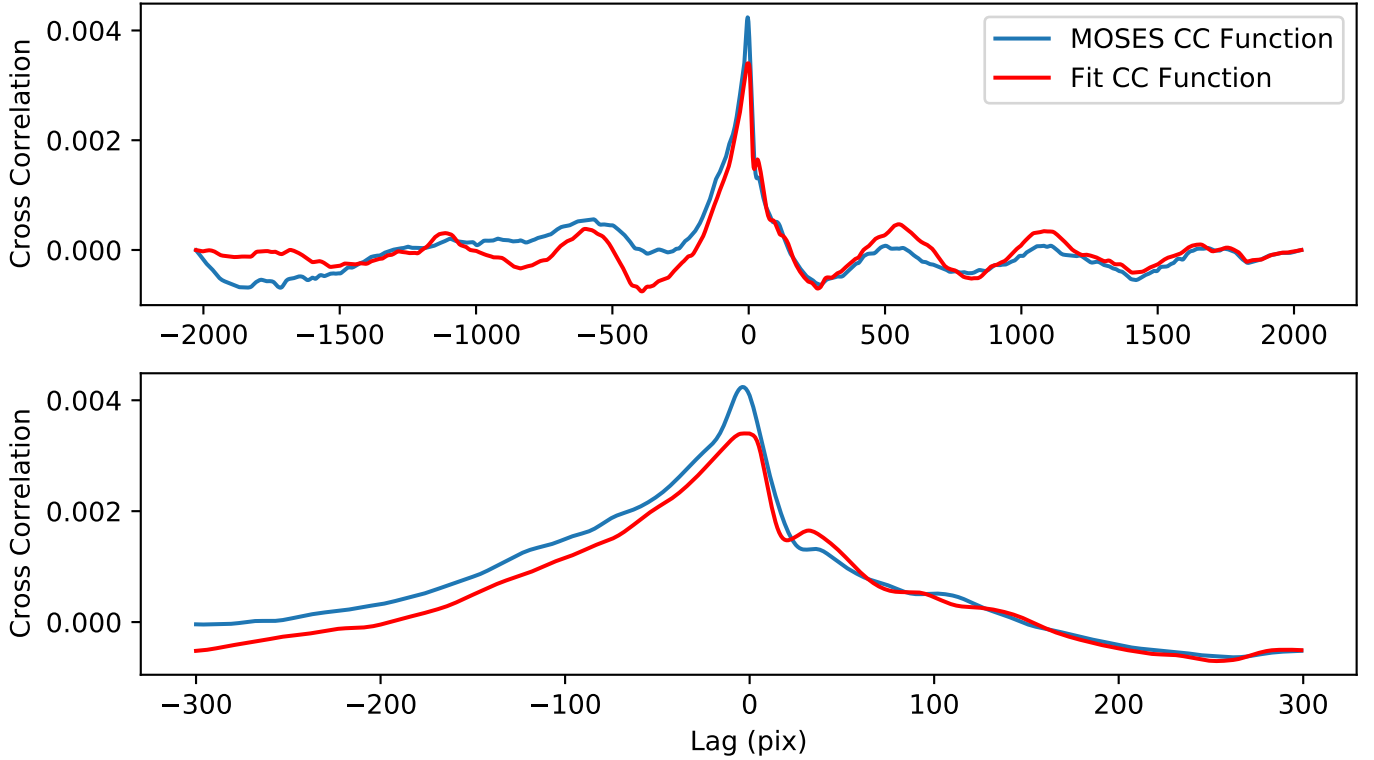


Figure 9. The top panel shows the entire mean cross-correlation along image rows of the MOSES (blue) and synthetic (red) difference images. The bottom panel shows a zoomed in view of the same function in order to highlight features near zero lag.

Table 1. Average MOSES Image Spectral Content

Ion	Contribution
He II	80.55%
Si XI	8.37%
Mg VIII	4.14%
Si VIII	2.65%
Si IX	1.33%
Fe XIII	1.29%
Other	1.67%

active region (region 2), are clearly visible and are adjacent to lighter smears of intensity to their left. Region 3, while not as sharp in the synthetic images, is reproduced at the limb with the same close separation between the positive and negative lobes.

Using the DEM of best fit we also generated an average spectrum for the MOSES data weighted by instrument's the wavelength dependent throughput (Figure 11). The synthetic spectrum of best fit was generated using the same constant pressure, abundance file, and spectral range as was used in the forward model (Section 3.3). MOSES images are dominated by the He II doublet near 303.7 Å. Si XI 303.3 Å, while much dimmer than He II, has the second highest intensity and represents approximately

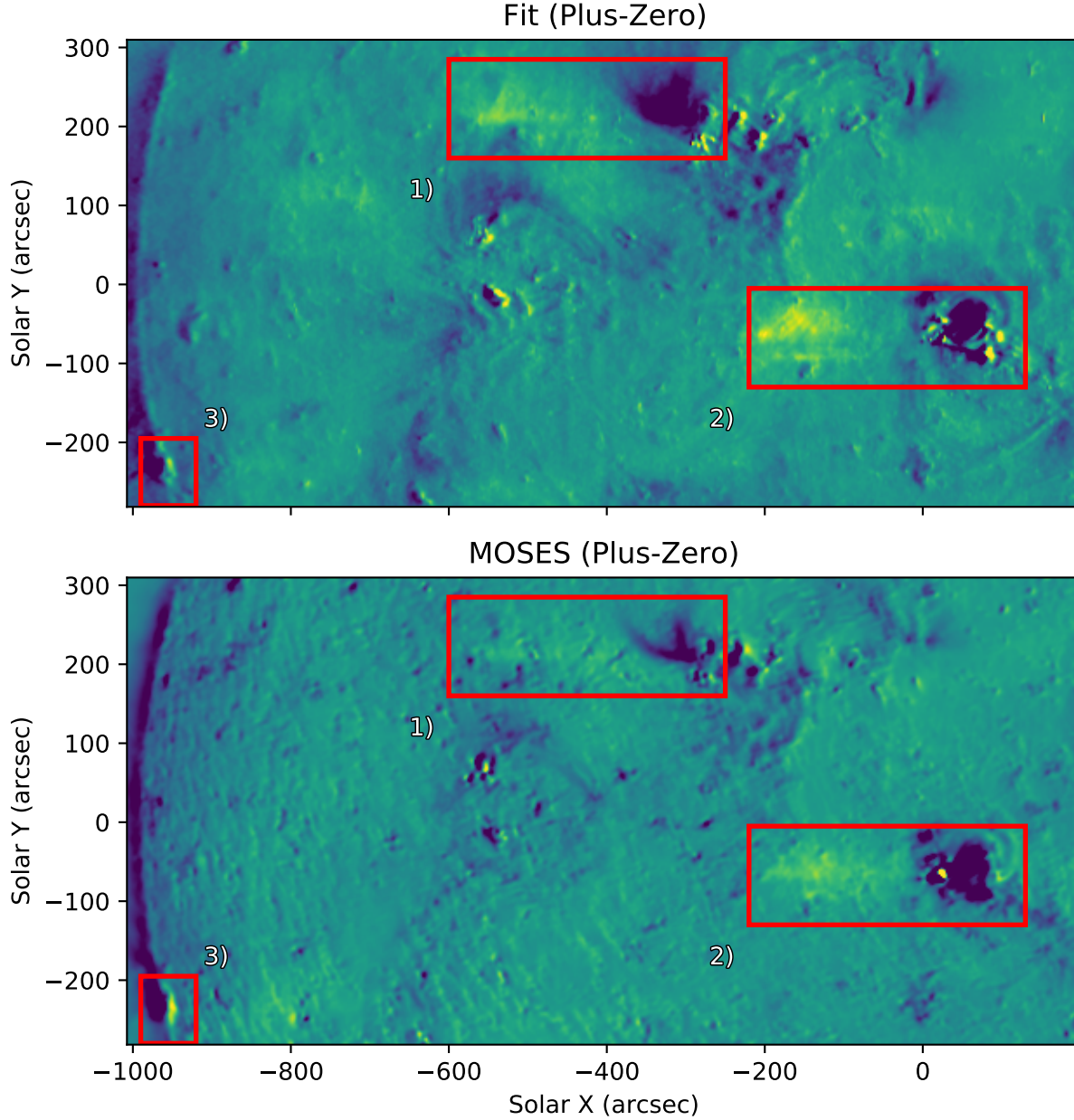


Figure 10. The $I_{+1} - I_0$ difference image, also shown in Figure 1c, along side the same difference image prepared from the synthetic MOSES images of best fit. Identical regions 1-3 are boxed in each image for comparison.

8 percent of the total intensity. The other notable, and unexpected contribution to the image intensity is from the host of lines between 310 and 320 angstroms, the brightest being from Mg VIII and Si VIII. Table 1 shows the total intensity contribution from each of the brightest ions.

5. DISCUSSION AND CONCLUSIONS

MOSES was successful in its goal of capturing solar images in the He II 303.8 Å emission line. Despite the use of a narrowband multi-layer coating on both the primary and secondary optics, MOSES captured several solar features not easily attributed to the dominant He II 303.8 Å or nearby

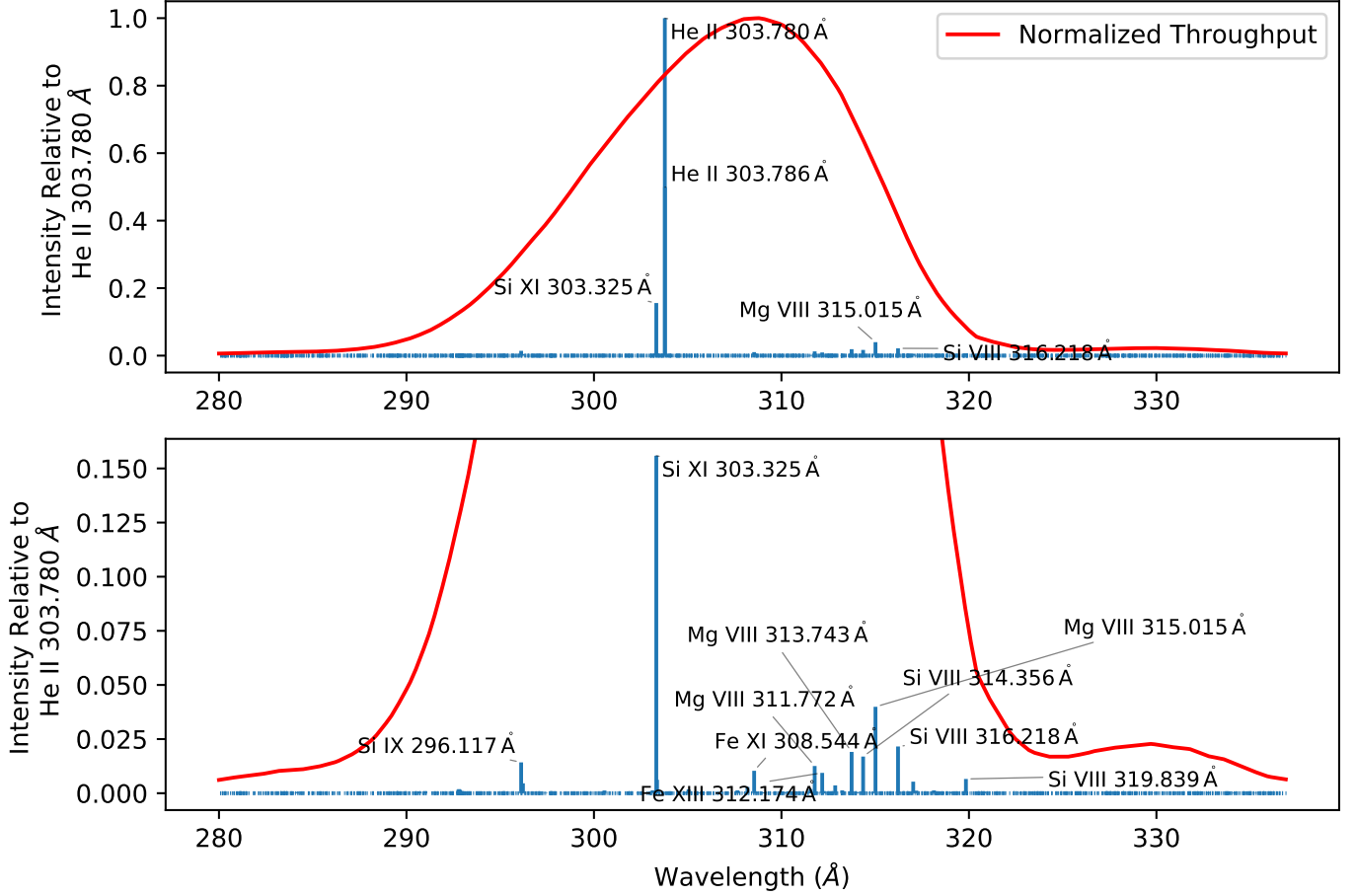


Figure 11. Using the DEM of best fit shown in Figure 8 we generated the above synthetic spectrum using Chianti. Each line intensity has also been weighted by the wavelength dependent throughput of the MOSES optical system (normalized and shown in red).

strong Si XI 303.3 Å lines. In order to identify and quantify the spectral content of these additional features we cross-correlated two MOSES difference images and identified peaks in correlation as significant and indicative of spectral contamination. Using a forward model that combines four co-temporal EIT images with synthetic spectra from Chianti we created synthetic MOSES difference images with known spectral content that could be cross-correlated and compared to the MOSES difference image cross-correlation function. By changing the DEM used we modified the spectral content of the synthetic MOSES data until we minimized the differences between the synthetic and real cross-correlation functions, the results of which are displayed in Section 4.

Compared to other solar DEMs, like those included in the Chianti software package, our DEM of best fit (Figure 8) shows a huge increase in emission measure between $\log T = 4\text{--}5$ K. This result is artificial and is caused by a significant lack of modeled intensity in the He II 303.8 Å emission line when using Chianti. The magnitude He II 303.8 Å intensity is incorrect by up to an order of magnitude when modeled under classical assumption and its lack of emission has been attributed to a host of factors including a lack of ionization equilibrium (Golding et al. 2017). Since the MOSES passband is absent of any other bright, cool lines (< 1 MK) our fit can increase the emission measure at low temperatures resulting in an increase in the intensity of He II 303.8 Å alone, allowing for a

more accurate intensity ratio between it and nearby hot emission lines like Si XI 303.3 Å. We find an average line ratio between He II 303.8 Å and Si XI 303.3 Å of $\approx 10:1$ which is inline with ratios measured by other slit spectrographs in regions of moderate solar activity (Cushman & Rense 1978).

It is also important to note that although we chose to use a DEM as the basis of our fit, we do not consider it to be an accurate average DEM of the sun across the MOSES field of view. The formation temperatures of lines within the MOSES passband very sparsely cover the range of $\log T = 4 - 6.5$, allowing multiple DEMs to represent the same spectral content. Despite the parameter space being very degenerate, we chose to vary emission measure rather than individual line intensities to ensure our final fits have physically realizable line ratios.

The temperature coverage of EIT, used to create the synthetic data, is also very sparse. Luckily, the dominant line in each of the four included EIT images have peak formation temperatures that closely match the peak formation temperatures of the brightest lines imaged by MOSES. The four EIT channels used, He II 304 Å, Fe IX 171 Å, Fe XII 195 Å, Fe XV 284 Å, have peak formation temperature of $\log T = 4.7$ K, 5.9 K, 6.2 K and 6.3 K respectively, calculated using Chianti. The bright, hotter lines imaged by MOSES, namely from Mg VIII, Si VIII, and Si XI, have peak formation temperatures of $\log T = 5.9$ K, 5.95 K, and 6.2 K respectively. This allows for synthetic MOSES data with key intensity contributors well represented, as is evident when comparing features in the synthetic difference images of best fit to those in the MOSES difference images (Figure 10).

The most mysterious features in the MOSES difference images, those that motivated this study, are regions 1 (the “wishbone”) and 2 identified in Figures 1 and 10. These two features are very coronal in appearance and easily identifiable in the hotter EIT channels, but they cannot be attributed to the most obvious source of contamination in any He II image, Si XI 303.3 Å, due to the lack of positive component ≈ 15.7 pixels away like that seen in region 3. The presence of a faint, light smear left of the wishbone indicates that it is from emission at longer wavelengths than He II 303.8 Å, but a lack of clear positive wishbone makes it difficult to attribute it to a single contaminant line. The best fit spectrum (Figure 11) shows a host of lines between 310 and 320 angstroms, mostly emitted by the Mg VIII and Si VIII ions, the strongest being Mg VIII 315 Å (Also identified by Rust (2017)). Despite their individually weak intensity, combined they contribute on order the same intensity as Si XI 303.3 Å. Due to spectral dispersion, these lines are very faint in the MOSES outboard order and could likely be neglected. In the zero order image, all of these faint lines land in the same place on the detector which is why these features appear so clearly when subtracting the zero order. The intensity contribution from lines other than He II 303.8 Å and Si XI 303.3 Å amounts to around ten percent of the total intensity in the zero order and comes from tens of lines that may be easily overlooked during analysis or instrument development. Some of these lines are shifted hundreds of pixels, causing feature within the field of view of the MOSES zero order to be shifted out and those outside to be shifted in the outboard orders. This results in different total intensities existing in each spectral order, which could lead too poor confusion when comparing orders during inversion.

While overlapping intensity from different spectral lines is inevitable in this type of instrument, the use of a zero order image and a lack of clearly defined field of view in MOSES makes it more difficult to deal with. ESIS (Smart et al. 2021; Parker et al. 2021), another sounding rocket based slitless spectrograph, builds off the MOSES concept and improves on it in several ways. The most relevant improvements include an intermediate focal plane and field stop, that clearly defines the instrument field of view, as well as a lack of zero order image. Including a field stop defines the

same field of view for every channel and solves the problem of highly shifted features existing in some orders, but not others. Even with a clearly defined field of view features in the zero order image would still have intensity from lines that are dispersed off the detector in the outboard orders. While the use of zero order image is attractive for a variety of reasons (co-alignment, capturing spatial information, etc.) the ambiguity of spectral content makes it difficult to compare its intensity to outboard orders without some type of forward modeling. Using all dispersed orders, as is done in ESIS, makes it much easier to compare the intensity of solar features from specific emission lines, and understand the possible errors in that intensity from overlapping lines. Future instruments of this kind considering using a zero order image should carefully consider the trade offs and do some modeling work to understand the impact of more subtle spectral lines on data analysis.

In summary, we have developed a forward model of the MOSES instrument that combines the Chianti database and images from EIT to form synthetic MOSES data with known spectral content. Our fits match the MOSES data well and indicate that the unexpected solar features identified in MOSES difference images are composed of tens of dim lines that contribute approximately 10 percent of the average intensity in the zero order image and are smeared over hundreds of pixels in the outboard orders. This unexpected spectral content makes it difficult to compare the intensity of certain features between spectral orders and, if not properly accounted for, could lead to a lack of convergence when inverting slitless spectrograph data.

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