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## **Coronal and Chromospheric Properties** of Flare-Imminent versus Flare-Quiet Active Regions

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#### INTRODUCTION

Solar flare forecasting has focused on characterizing an active region's complexity using white-light images, photospheric field maps (and extrapolations using those maps as boundary conditions), and chromospheric morphology. We report on the results of a large-sample study of "AIA Active-Region Patches" (AARP) regions that uses NonParametric Discriminant Analysis through the NWRA Classification Infrastructure (NCI) to ask the question of what distinguishes flare-imminent vs. flare-quiet active regions (or epochs during a region's evolution). Parametrizing using area-summations and high-order moment analysis on both direct and running-difference images in multiple AIA wavelength bands, we find top Brier Skill Scores in the 0.07–0.33 range, True Skill Statistics in the 0.68–0.82 range (depending on event definition), and Receiver Operating Characteristic Skill Scores above 0.8. We find evidence for enhanced heating and kinematics in flare-imminent samples. Coronal- and chromospheric- targeted parameters, like the commonly-used photosphere-magnetic field parameters, are candidates for improved physics-based understanding of the flare-ready atmosphere, and improved flare forecasting.

#### **AARP DATABASE**

AIA Active Region Patches (AARPs) [3] are region-targeted extractions of AIA full-disk data analogous to HMI Active Region Patches (HARPs) that are defined using the line-of-sight photospheric field [1]. AARPs are extracted from EUV images in 94, 131, 171, 193, 211, 304, 335 Å covering the chromosphere to corona, sensitive to emission from temperatures ranging from 0.5 to 10 MK, plus the UV 1600 Å sensitive to upper photosphere and transition region emission at 0.1 MK. Temporal downsampling consists of bursts of 13 minutes of data sampled at 72s, hourly for 7 hours centered at :48 (15:48–21:48 UT). The AARP dataset is coordinated with data previously produced for photospheric active region investigations using the vector photospheric field [6].

Date Range HARP Range NOAA AR Range # Samples # AARP days Archive

#### **EXAMPLE NONPARAMETRIC PROBABILITY DENSITY FUNCTIONS**







Overview of the AARP extraction and down-sampling in the spatial and temporal domains: HARP/AARP 1447 on March 07, 2012. All AARP data (FITS files) will be available through the NASA/Solar Data Analysis Center.

## PARAMETRIZATION OF CHROMOSPHERIC, TRANSITION REGION, AND CORONAL BEHAVIOR

"For solar active regions, are flare-imminent epochs distinguishable from flare-quiet epochs on the basis of chromospheric and coronal emission and kinematics?" We ask this using UV and EUV intensity images and HMI-defined active regions, *without* the benefit of spectroscopy [9], but *with* timeseries analysis [2] in order to enhance physical interpretation of the results. Moment analysis, including higher-order moments, describes bulk properties and the more behavior of the distribution wings respectively. The parametrization uses the "direct" images ("I"), and the running-difference images  $(`'\Delta I'')$ :

- The total brightness  $\Sigma(I_*)$ , and total running-difference  $\Sigma(\Delta I_*)$ , are extensive parameters that scale with size of the target.
- The moments of the brightness distribution  $\mathcal{M}(I_*)$ : the mean  $\mu(I_*)$ , standard deviation  $\sigma(I_*)$ , skew  $\varsigma(I_*)$ , and kurtosis  $\kappa(I_*)$ ; moments are intensive parameters that do not scale with target size. Similarly, the moments of the running-difference image distributions  $\mathcal{M}(\Delta I_*)$ . • For each of the 80 base parameters we derive a "static" state and its temporal behavior dX/dt, using the slope and intercept of a linear fit, for a final tally of 160 parameters evaluated. • Note: The  $\Delta I_*$  parameters summarize high-frequency variation; the dX/dt parameters summarize the longerterm trends in behavior, including that of high-frequency variations. • These parametrizations are chosen to be physically interpretable. For example the appearance of new bright loops will enhance overall brightness levels of, e.g., 171Å images ( $\Sigma(I_{171})$ ) and the mean brightness levels ( $\mu(I_{171})$ ), but also produce a positive skew in the associated running-difference images ( $\varsigma(\Delta I_{171})$ ) as the new loops appear.

NonParametric Probability Density Functions of four parameters for the C1.0+/24 hr event definition: (left-toright): the total of the 94Å emission ( $\ln \Sigma(I_{94})$ ), the mean of the 94Å emission ( $\ln \mu(I_{94})$ ), the kurtosis of the runningdifference of 304Å images (ln  $\kappa(\Delta I_{304})$ ), and the change with time of the skew of the 211Å emission ( $d\varsigma(I_{211})/dt$ ). For all, the **event**, **non-event** non-parametric density estimates using an adaptive kernel are shown, their means (---/--), and the discriminant boundary(ies).

### **CORONAL & CHROMOSPHERIC INDICATORS OF** FLARE-IMMINENT VS. FLARE-QUIET ACTIVE REGION EPOCHS



**Top:** Reliability plots for (left:right) C1.0+/24 hr, C1.0+/6 hr, M1.0+/24 hr, M1.0+/6 hr for high-performing parameters according to the Brier Skill Scores; the x = y line indicates perfect reliability, the blue histogram is the frequency of occurrence for each prediction bin, the light-blue horizontal line indicates the climatology and the red dashes indicate the "no skill" line. Error bars reflect the sample size in each bin. **Bottom:** Receiver Operating



Steps for parametrizing the AARPs. The direct and running-difference images of 304Å are shown (left two panels); the kurtosis of the running-difference images is calculated for the 13min (11 images) (" $\kappa(\Delta I_{304})$ ", mid-right,  $\Box$ ) from which the mean and standard deviation are shown (\*  $\pm$ ); these become the data points for each of the samples covering 7 hours (right,  $\Box \pm$ ), from which the linear slope and last-data (21:48 UT) intercept (\*) provide the final inputs for nonparametric discriminant analysis. Shown: a "no-event" epoch @2011.07.25 and "yes-event" epoch @2011.07.26.

# NWRA CLASSIFICATION **INFRASTRUCTURE (NCI)**

• Well-established statistical classifier system based on Nonparametric Discriminant Analysis (NPDA) [6]. • Four components at work in this facility: the input parameters, the event definitions and event lists, the statistical package, and the evaluation metrics. • NPDA: functional form of the distributions is not assumed, but estimated from the data.

<b>EVENT DEFINITIONS</b>
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Label GOES lower Validity Latency # Events,								
	limit	Period	Period	Event				
	$(10^{-6} \mathrm{Wm}^{-2})$	(hr)	(hr)	Rate				
C1.0+/24 h	r 1.0	24	0.2	2752 (0.086)				
M1.0+/24 h	r 10.0	24	0.2	450 (0.014)				
C1.0+/6 hr	1.0	6	0.2	1262 (0.039)				
M1.0+/6hr	10.0	6	0.2	155 (0.005)				

Characteristic (ROC) plots for (left:right) C1.0+/24 hr, C1.0+/6 hr, M1.0+/24 hr, M1.0+/6 hr, for the parameters as indicated, generally in the top-10 or bottom-5 scoring group.



Radar plots showing Brier Skill Score for all parameters, grouped by filter, as labeled. Arcs indicate the range of the BSS  $\pm \sigma$ BSS, BSS > 0 (blue) and |BSS| for BSS < 0 (orange), with darker hues indicating the uncertainty ranges. Shown: C1.0+/24 hr event definition results for two filters 94Å and 131Å (left) and M1.0+/24 hr event definition results for the same two filters (right).

### SAMPLE SKILL SCORES & METRICS

Parameter	Brier	Max(TSS)	Gini	Parameter	Brier	Max(TSS)	Gini	
	Skill Score		(ROCSS)		Skill Score		(ROCSS)	
Top 10 S	coring Parame	ters: C1.0+/	24 hr	Top 10 Scoring Parameters: M1.0+/24 hr				
$\kappa(\Delta I_{94})$	$0.332 \pm 0.011$	0.650	0.816	$\kappa(\Delta I_{94})$	$0.160 \pm 0.015$	0.794*	0.909*	
$\kappa(\Delta I_{131})$	$0.315\pm0.011$	0.658	0.810	$\kappa(\Delta I_{131})$	$0.132\pm0.010$	0.734	0.862	
$\kappa(\Delta I_{171})$	$0.312\pm0.012$	0.670	0.809	$\Sigma(\Delta I_{94})$	$0.131 \pm 0.015$	0.704	0.840	
$\kappa(\Delta I_{304})$	$0.310\pm0.010$	0.668	0.812	$d\kappa(\Delta I_{94})/dt$	$0.125 \pm 0.018$	0.680	0.837	
$\Sigma(I_{94})$	$0.302\pm0.013$	0.680*	0.830*	$\Sigma(\Delta I_{131})$	$0.118 \pm 0.021$	0.640	0.786	
$\kappa(\Delta I_{193})$	$0.301\pm0.011$	0.657	0.814	$\kappa(\Delta I_{211})$	$0.117 \pm 0.008$	0.750	0.863	
$\kappa(\Delta I_{211})$	$0.291 \pm 0.011$	0.651	0.794	$d\varsigma(\Delta I_{94})/dt$	$0.116 \pm 0.009$	0.640	0.802	
$\Sigma(\Delta I_{94})$	$0.286 \pm 0.011$	0.626	0.788	$\kappa(\Delta I_{304})$	$0.116 \pm 0.010$	0.725	0.851	
$\Sigma(I_{335})$	$0.280 \pm 0.014$	0.672	0.822	$d\varsigma(\Delta I_{131})/dt$	$0.110\pm0.016$	0.658	0.812	
$d\Sigma(\Delta I_{94})/dt$	$0.273 \pm 0.011$	0.597	0.761	$d\Sigma(\Delta I_{131})/dt$	$0.109 \pm 0.017$	0.626	0.764	
Bottom 5 Scoring Parameters: C1.0+/24 hr				Bottom 5 Scoring Parameters: M1.0+/24 hr				
$d\mu(I_{193})/dt$	$0.001 \pm 0.000$	0.035	0.046	$d\sigma(I_{171})/dt$	$0.000 \pm 0.000$	0.049	0.038	
$\mu(\Delta I_{171})$	$0.001\pm0.001$	0.052	0.047	$d\mu(I_{171})/dt$	$0.000\pm0.000$	0.028	0.019	
$d\sigma(I_{171})/dt$	$0.000\pm0.000$	0.023	0.015	$d\mu(\Delta I_{1600})/dt$	$0.000 \pm 0.000$	-0.031*	-0.036*	
$d\kappa(I_{131})/dt$	$-0.011 \pm 0.009$	0.203	0.283	$\mu(\Delta I_{1600})$	$0.000\pm0.000$	-0.031	-0.024	
$d\kappa(I_{335})/dt$	$-0.068 \pm 0.028$	0.160	0.262	$d\mu(I_{131})/dt$	$-0.002 \pm 0.004$	0.157	0.197	

- New capabilities: adaptive kernel for density estimates, multiple population (as well as multiple parameter) discriminant analysis, region-based crossvalidation.
- Bootstrap with replacement [4, 5, 6], also region-based with-holding, provides uncertainties on the evaluation metrics.

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#### **THE TAKE-AWAY POINTS:**

• Total emission can identify flare-ready regions, but mean brightness measures do not: this is the wellknown active-region-size/flare-productivity relation.

• Once a region is flare productive, the active-region coronal plasma appears to stay hot.

• 94 Å provides the most parameters with discriminating power; it probably benefits from sampling mul-

- tiple physical regimes.
- Flare-imminent regions preferentially display shortlived small-scale brightening events.
- Longer-term evolution and trends can provide flareimminent information, but at no strong preference over "static" parameters.
- Improvements to flare forecasts will likely come by combining coronal and photospheric data.

\*: Top or Bottom score for Max(TSS) and for  $\mathcal{G}$ .

For C1.0+/24 hr the worst Max(TSS)=-0.038, and G=-0.034 both for  $d_{S}(I_{211})/dt$  which has BSS=0.001 ± 0.001.

#### **INTERPRETATION AND ADDITIONAL POINTS**

• *Running-Difference images dominate the top-scoring parameters*. The short-duration fluctuations or temporal changes are overall important to consider, in addition to "direct" images of the coronal & transition region. • *The fluctuations are small.* Higher-order moments of running-difference images are most influenced by spatially small-scale but intense variations. Spatially binning the AIA images will cause these characteristics to be lost. • Longer-term evolution dominates in the bottom-scoring parameters. The temporal variation of the image brightness distributions over a few hours generally do not well-separate the flare-ready samples. • Solar cycle makes very little difference. Cycle-related event-rate variations are more impactful to classification success

than, *e.g.* background emission variations. (Details in Leka+2023).