

SOLARNET: SOLAR FLARES PREDICTION WITH SELF-SUPERVISED LEARNING

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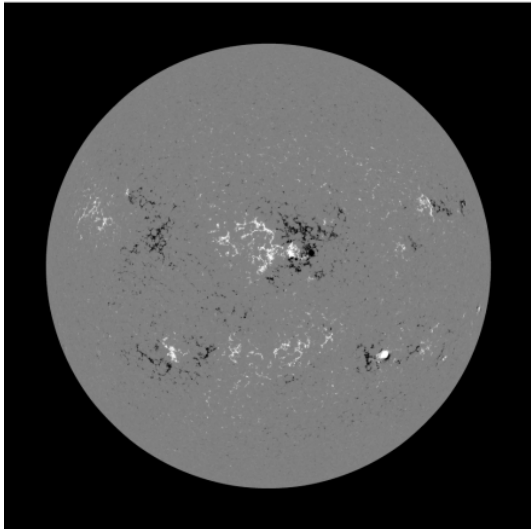
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March 21, 2022, Boulder CO

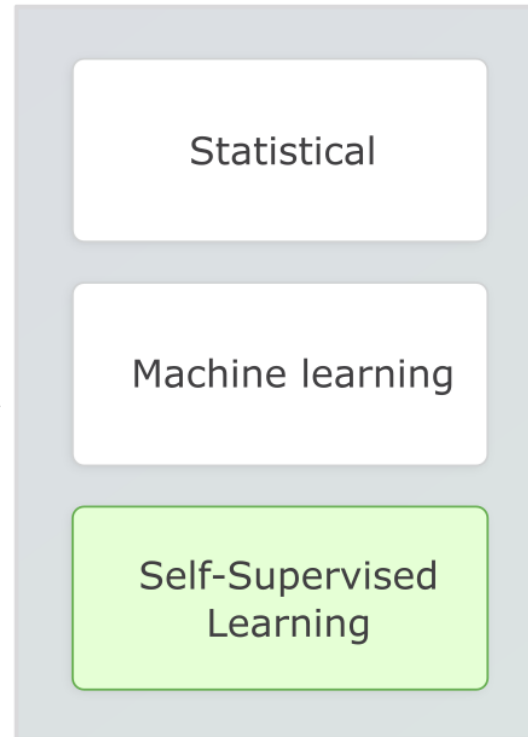
CONTEXT

Input data



Sun magnetograms from SDO/HMI

Model



Quiet sun

\geq C-class flare

Predictions or probabilities for solar flares

STATE OF THE ART

Reference	Method	Data	True Skill Statistic (TSS)	
			≥C-class	≥M-class
Park et al. (2018)	CNN	SOHO/MDI, SDO/HMI	0,63	
Nishizuka et al. (2018)	MLP	SDO/HMI&AIA	0,63	0,8
Wang et al. (2019)	LSTM	SHARP	0,559	0,681
Ahmed et al. (2013)	CCNN	SOHO/MDI	0,512	
Huang et al. (2018)	CNN	SOHO/MDI, SDO/HMI AR	0,487	0,662
Bloomfield et al. (2012)	Poisson prob.	Flares & AR	0,456	
Jonas et al. (2017)	Linear classifier	SDO/HMI&AIA AR		0,814
Bobra and Couvidat (2015)	SVM	SHARP		0,761
Zheng, Li, and Wang (2019)	CNN	SDO/HMI AR		0,749
Liu et al. (2017)	Random forest	SHARP		0,532

CONTRIBUTIONS

1. **Self-Supervised Learning (SSL)** applied to solar images
2. Prepared **datasets** for SSL pre-training and flares prediction
3. Seamless **reproducibility** of results with a **library** to use the pre-trained models

SELF-SUPERVISED LEARNING

Step 1

Pretraining with unlabelled data

Pretext task

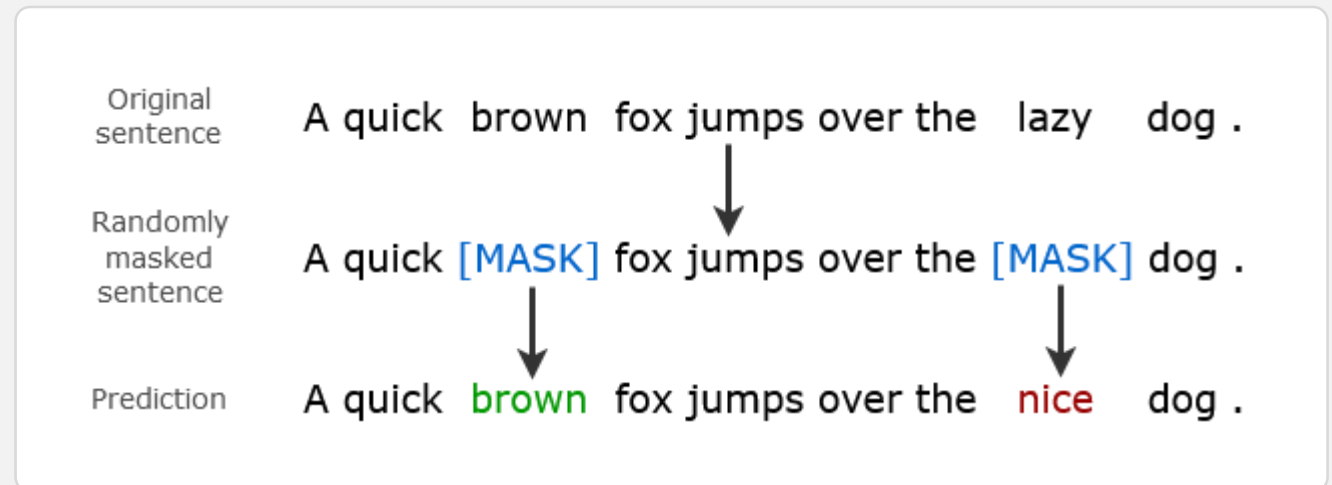
Large volume of data

➤ Learns pattern in the data

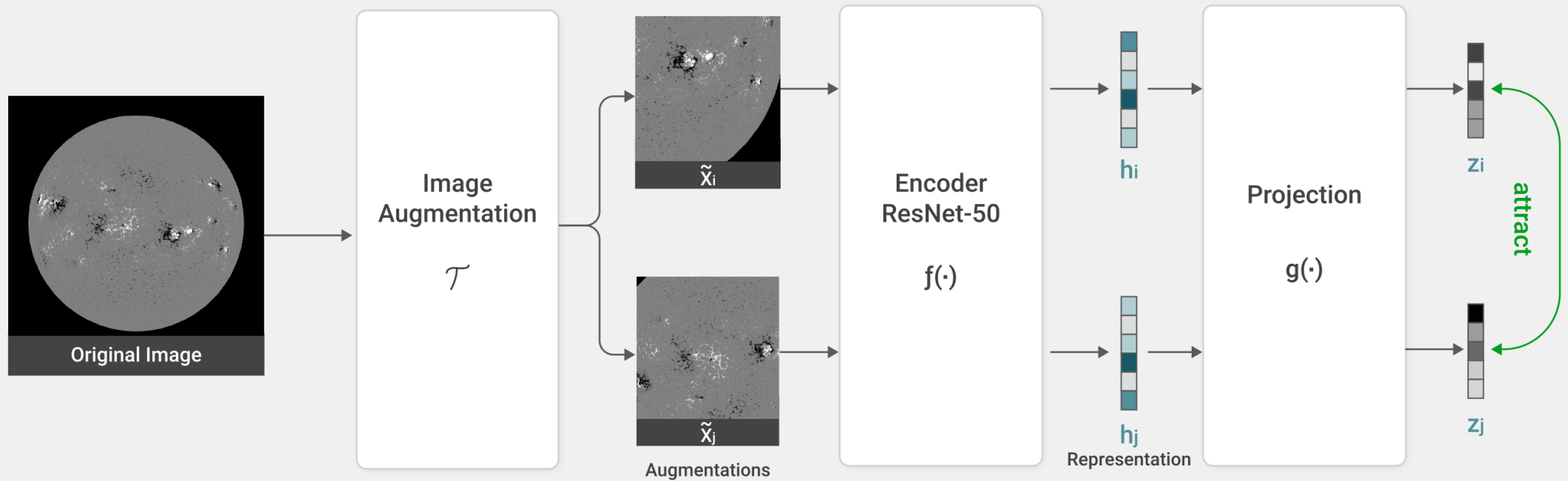
➤ General-purpose model

Step 2

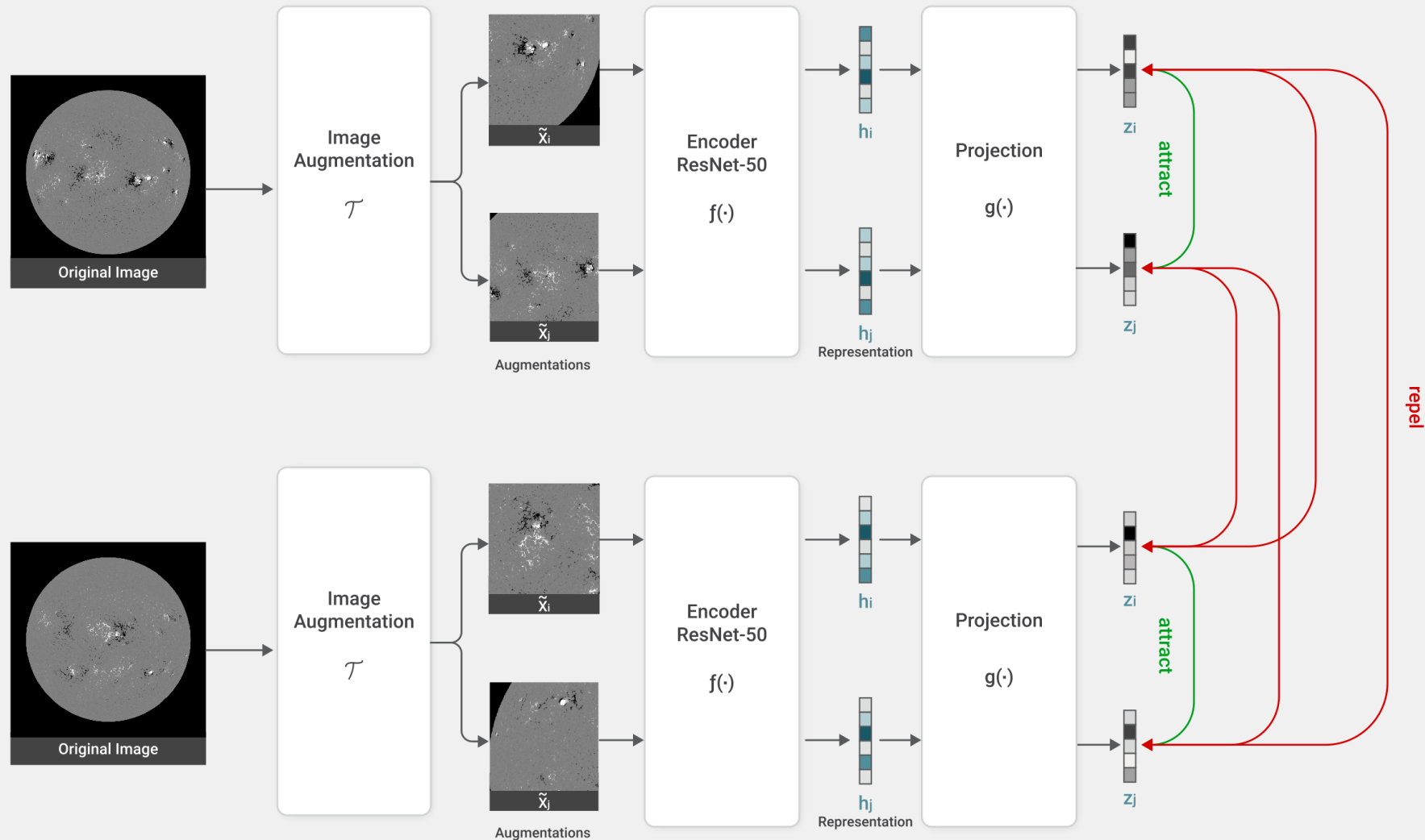
Finetuning with lower amount of data



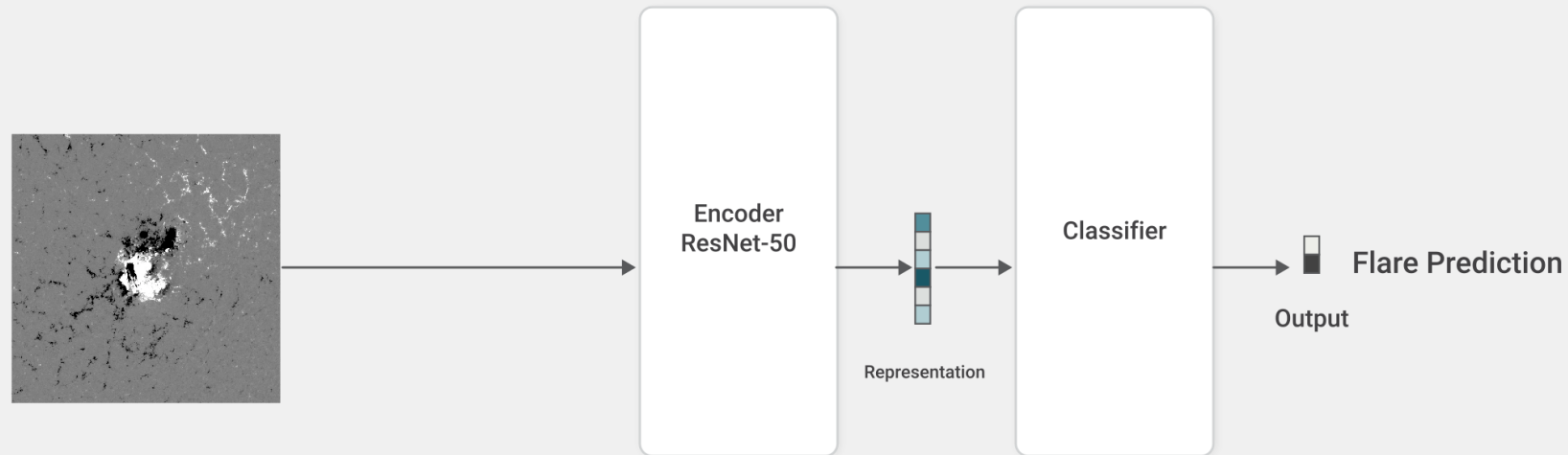
Step 1: Self-Supervised Learning



Step 1: Self-Supervised Learning



Step 2: Finetuning on supervised downstream task



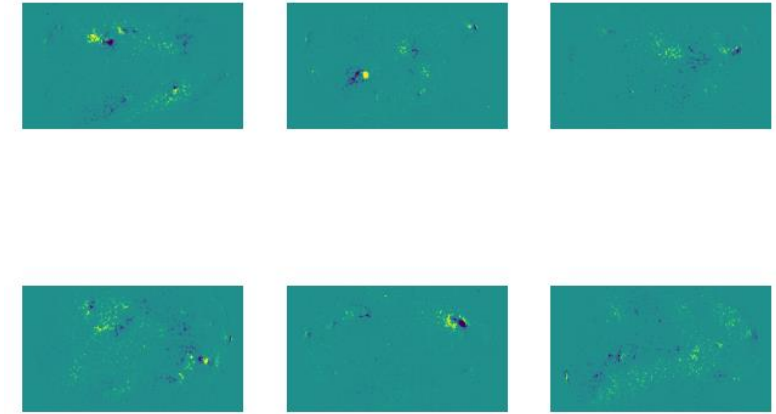
DATASETS

■ SDO-Dataset

from “A Machine Learning Dataset Prepared From the
NASA Solar Dynamics Observatory Mission”

Full-disk images 512x512, corrected
2010-2018

Step 1: Pre-training

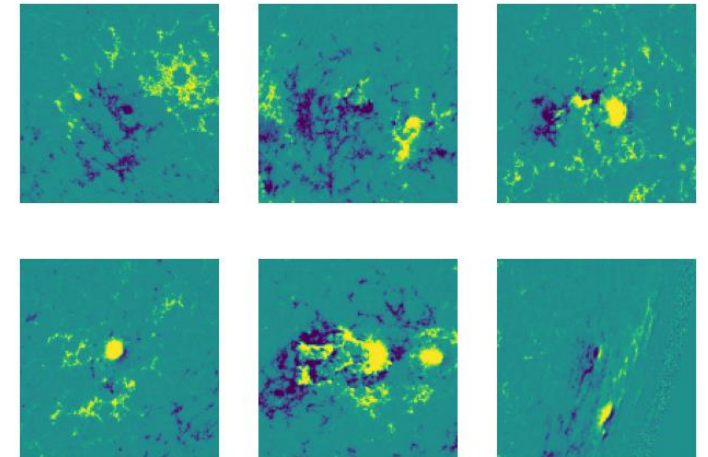


■ SDOBenchmark

images of active regions 256x256, preprocessed

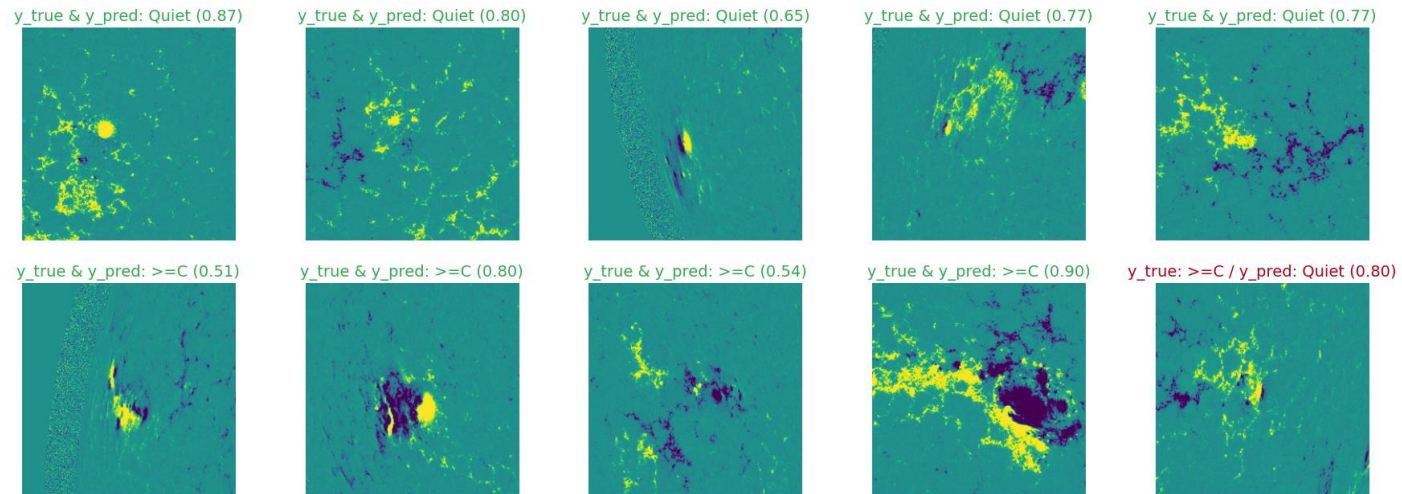
Labeled with peak flux
2012-2018

Step 2: Finetuning



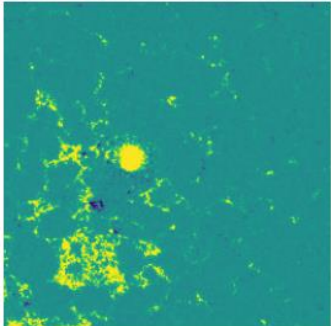
RESULTS

Dataset	Model	Split	TSS	F1
	Naive (random)		0.0	0.5
SDOBenchmark (active regions)	CNN (baseline)	Train/test split: SDOBenchmark	0.607	0.803
	ResNet-50 with SSL	Pretrain: SDO-Dataset 2010-2011 Train/test split: SDOBenchmark	0.646	0.819
Predict “Quiet sun” or “Flare \geq C-class” in next 24 hours, Data is magnetogram				

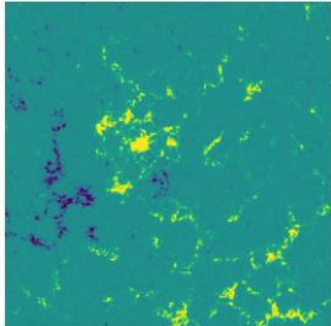


RESULTS

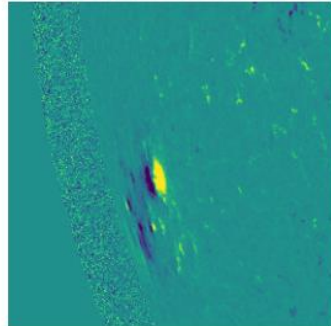
y_true & y_pred: Quiet (0.87)



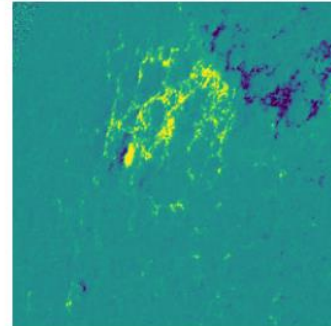
y_true & y_pred: Quiet (0.80)



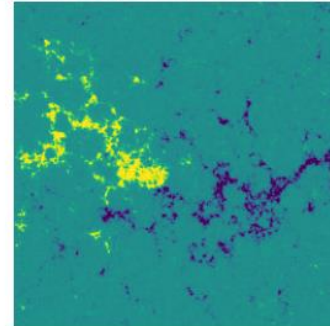
y_true & y_pred: Quiet (0.65)



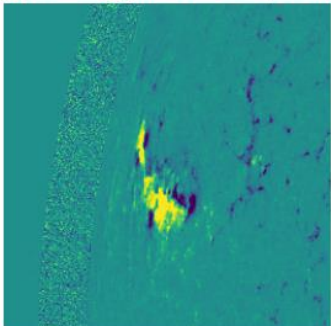
y_true & y_pred: Quiet (0.77)



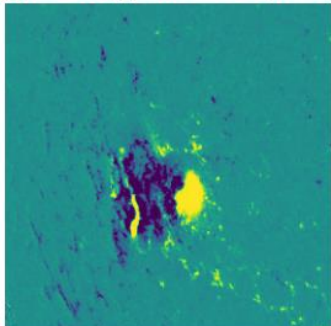
y_true & y_pred: Quiet (0.77)



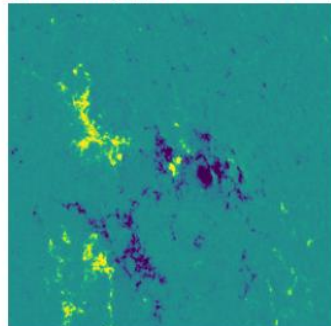
y_true & y_pred: $\geq C$ (0.51)



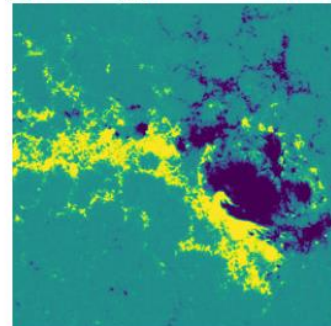
y_true & y_pred: $\geq C$ (0.80)



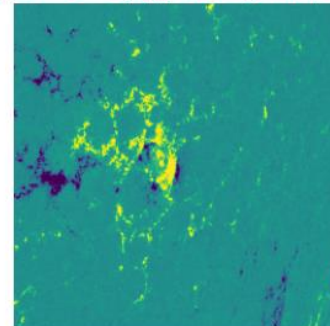
y_true & y_pred: $\geq C$ (0.54)



y_true & y_pred: $\geq C$ (0.90)



y_true: $\geq C$ / y_pred: Quiet (0.80)



RESULTS

Dataset	Model	Split	TSS	F1
	Naive (random)	-	0.0	0.5
SDO-Dataset (full-disk)	CNN (baseline)	Operational case: Train: 2010-2014 Test: 2015-2017	0.404	0.707
	ResNet-50 with SSL		0.446	0.726
	ResNet-50 with SSL	Split by months: Train: Jan.-Nov. Test: Dec.	0.783	0.891
Predict “Quiet sun” or “Flare \geq C-class” in next 24 hours, Data is magnetogram				

REPRODUCIBILITY & LIBRARY

Links to repositories

- Reproducible results:
gitlab.com/jdonzallaz/solarnet-thesis
- Library:
gitlab.com/jdonzallaz/solarnet

Get started

1. `>> pip install solarnet-lib`

2. `model = ImageClassification.from_pretrained("solarnet-ssl-bz-ft-month")`
`with torch.no_grad():`
`y_pred = model(tensor)`

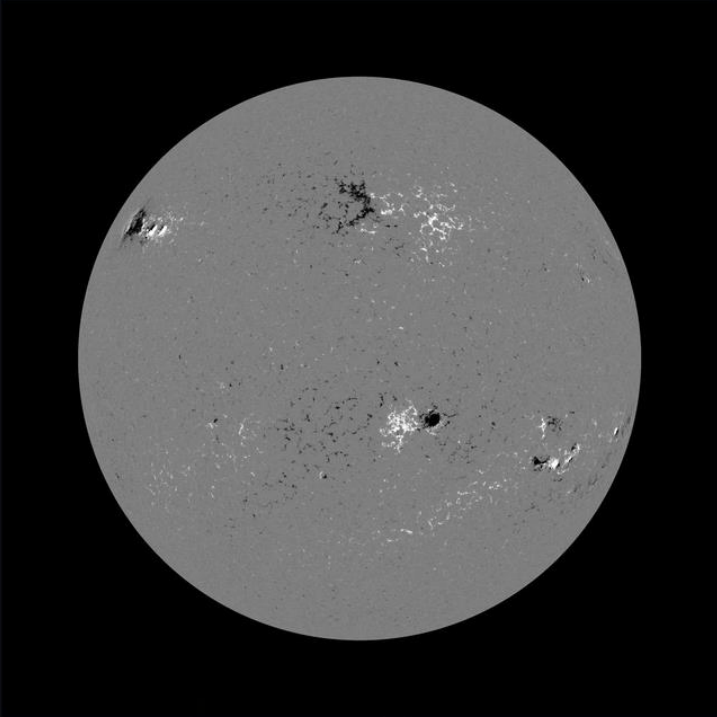
WORKING WITH REAL-TIME DATA

SolarNet

Online flare forecasting

Observation date

Checking prediction at closest date: 3/10/2022 , 12:16:57 PM



Prediction for next 24 hours
≥C-class flare
Confidence: 0.74

The Sun magnetogram at 3/10/2022, 12:16:57 PM

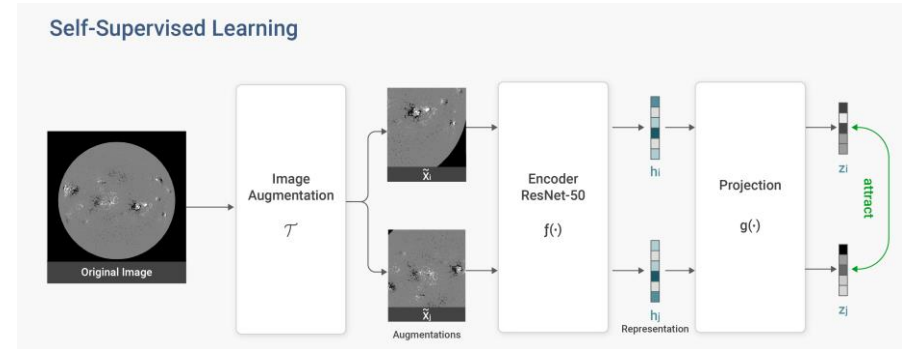
solarnet.kube.isc.heia-fr.ch

POTENTIAL DEVELOPMENTS

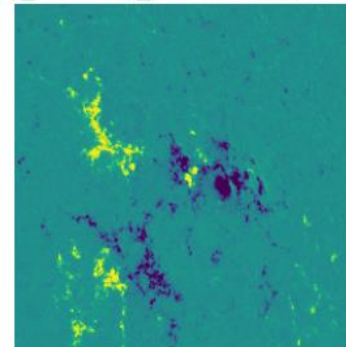
- Use of multi-channel images (see Yana Shtyk's poster)
- Evaluate other architectures & SSL approaches
- In-depth operational testing
- Extend pre-training dataset to 1996-2022

SUMMING UP

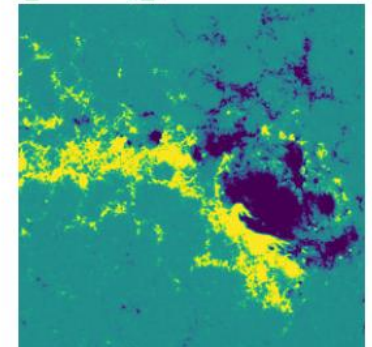
- Self-Supervised Learning – provides good representation of data
- Consistently improving results
- Great reproducibility and reusability



y_true & y_pred: >=C (0.54)



y_true & y_pred: >=C (0.90)



solarnet.kube.isc.heia-fr.ch

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LINKS

Library: gitlab.com/jdonzallaz/solarnet

Docs: jdonzallaz.gitlab.io/solarnet

Application: solarnet.kube.isc.heia-fr.ch

REFERENCES

Source of images: SDOML, SDOBenchmark & Helioviewer