



Deep Learning: Teaming up Machines and Humans to Peek into the Black Box

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Southwest Research Institute
High Altitude Observatory
National Solar Observatory

What is deep learning?



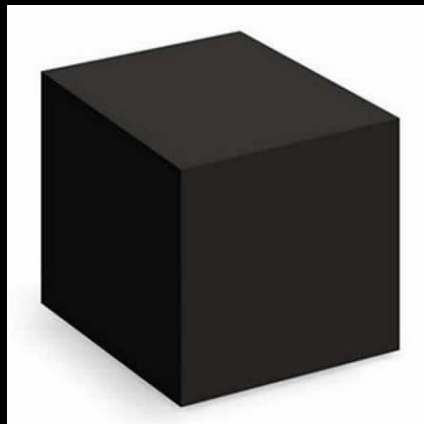
What techies think it is



What advocates think it is



What it really is



What skeptics think it is



What the public thinks it is

What is deep learning?



What techies think it is

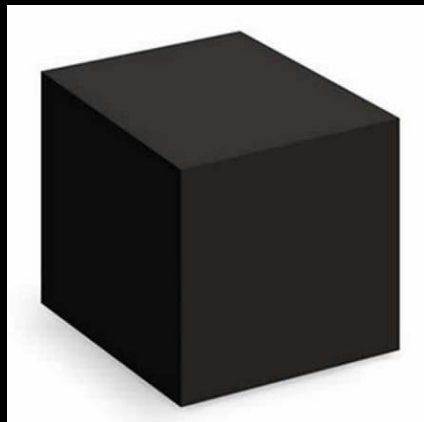


What advocates think it is

Getting past preconceptions is a big challenge



What it really is



What skeptics think it is



What the public thinks it is



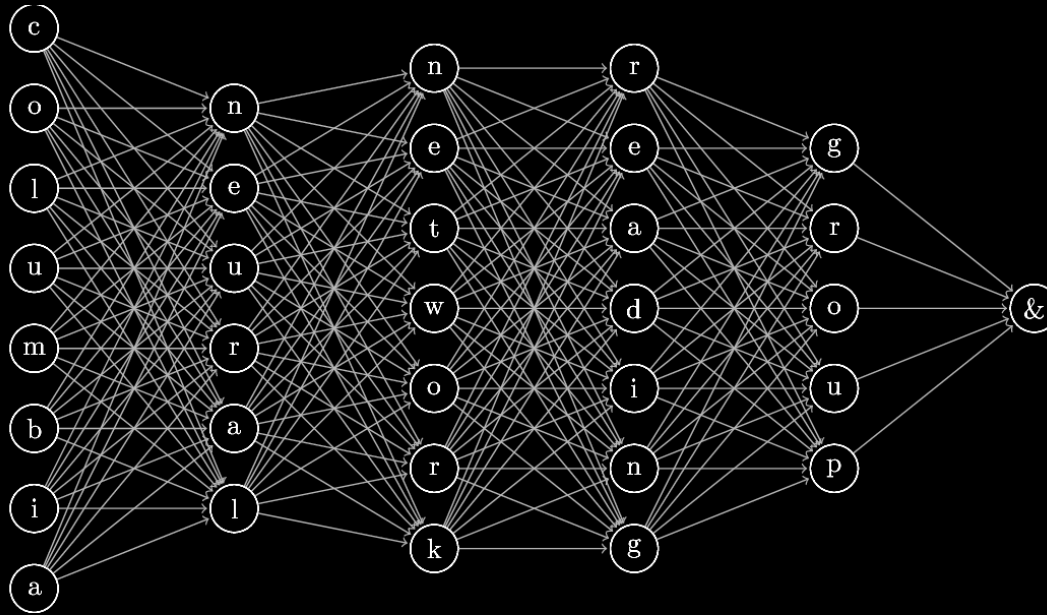
A tool that helps us find new things in our data

What is deep learning?*

A class of machine learning algorithms that:

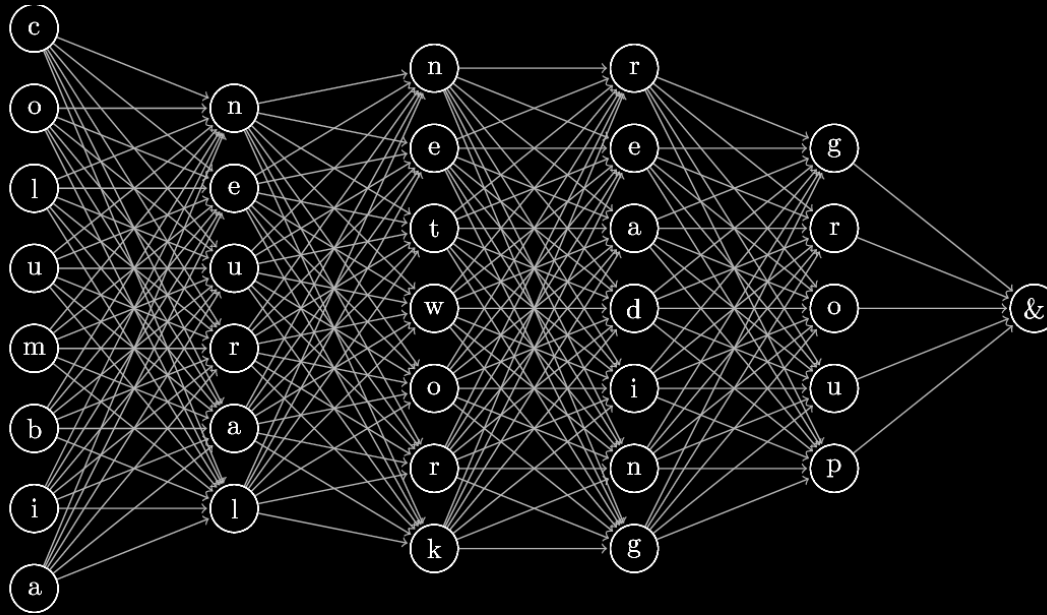
- Use a cascade of multiple layers of nonlinear processing units for feature extraction and transformation.
- Learn multiple levels of representations that correspond to different levels of abstraction (i.e. the levels form a hierarchy of concepts).

What is deep learning?*



- Learn multiple levels of representations that correspond to different levels of abstraction (i.e. the levels form a hierarchy of concepts).

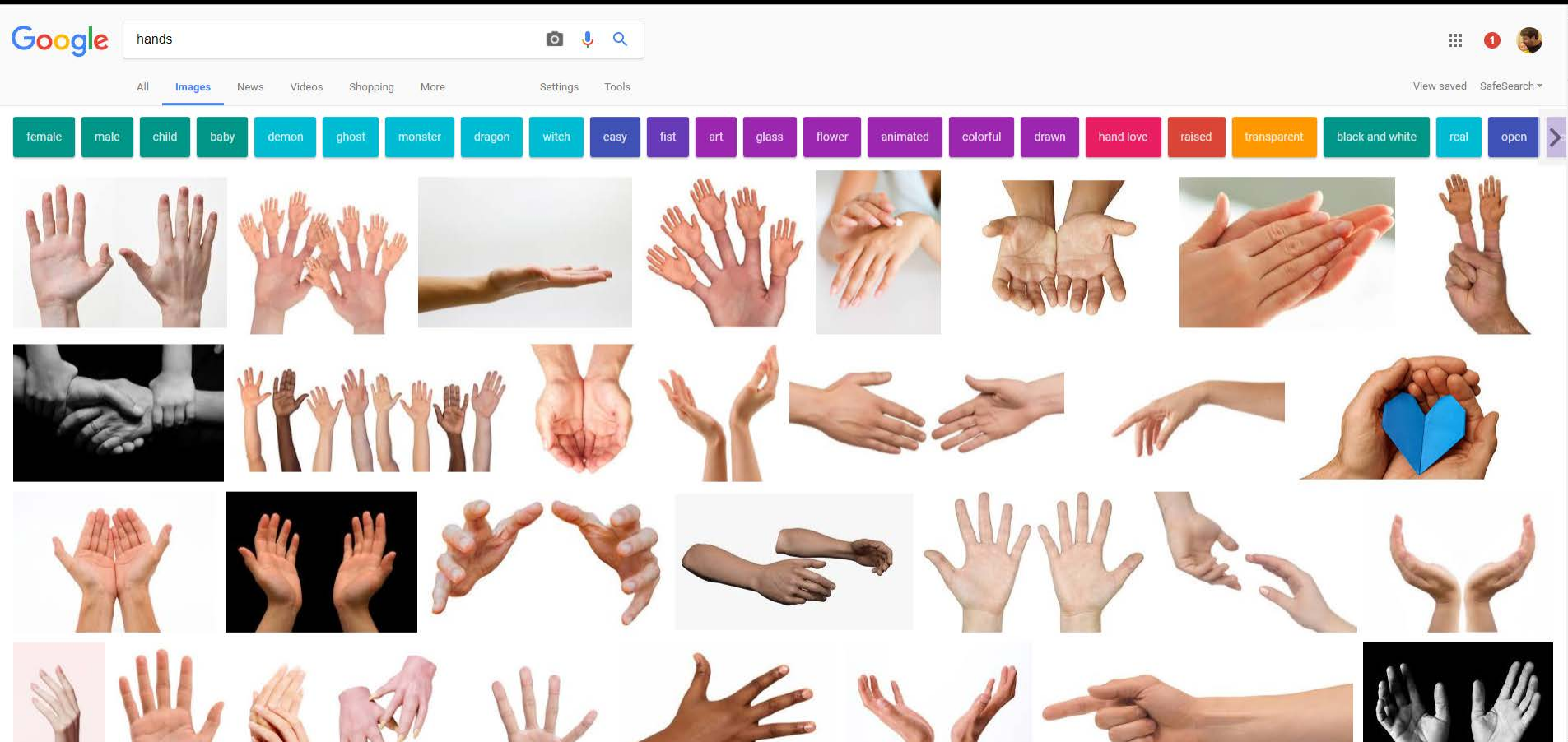
What is deep learning?*



Neural networks are not black boxes.

Why deep learning?

Deep learning has Important limitations too



Deep learning has important limitations too

TL;DR \ MICROSOFT \ WEB \

Twitter taught Microsoft's AI chatbot to be a racist asshole in less than a day

by James Vincent | @jvincent | Mar 24, 2016, 6:43am EDT

f SHARE t TWEET in LINKEDIN



It took less than 24 hours for Twitter to corrupt an innocent AI chatbot. Yesterday, Microsoft [unveiled Tay](#) — a Twitter bot that the company described as an experiment in "conversational understanding." The more you chat with Tay, said Microsoft, the smarter it gets, learning to engage people through "casual and playful conversation."

NOW TRENDING



The Pixel 2 XL drama is undermining Google's entire Pixel project

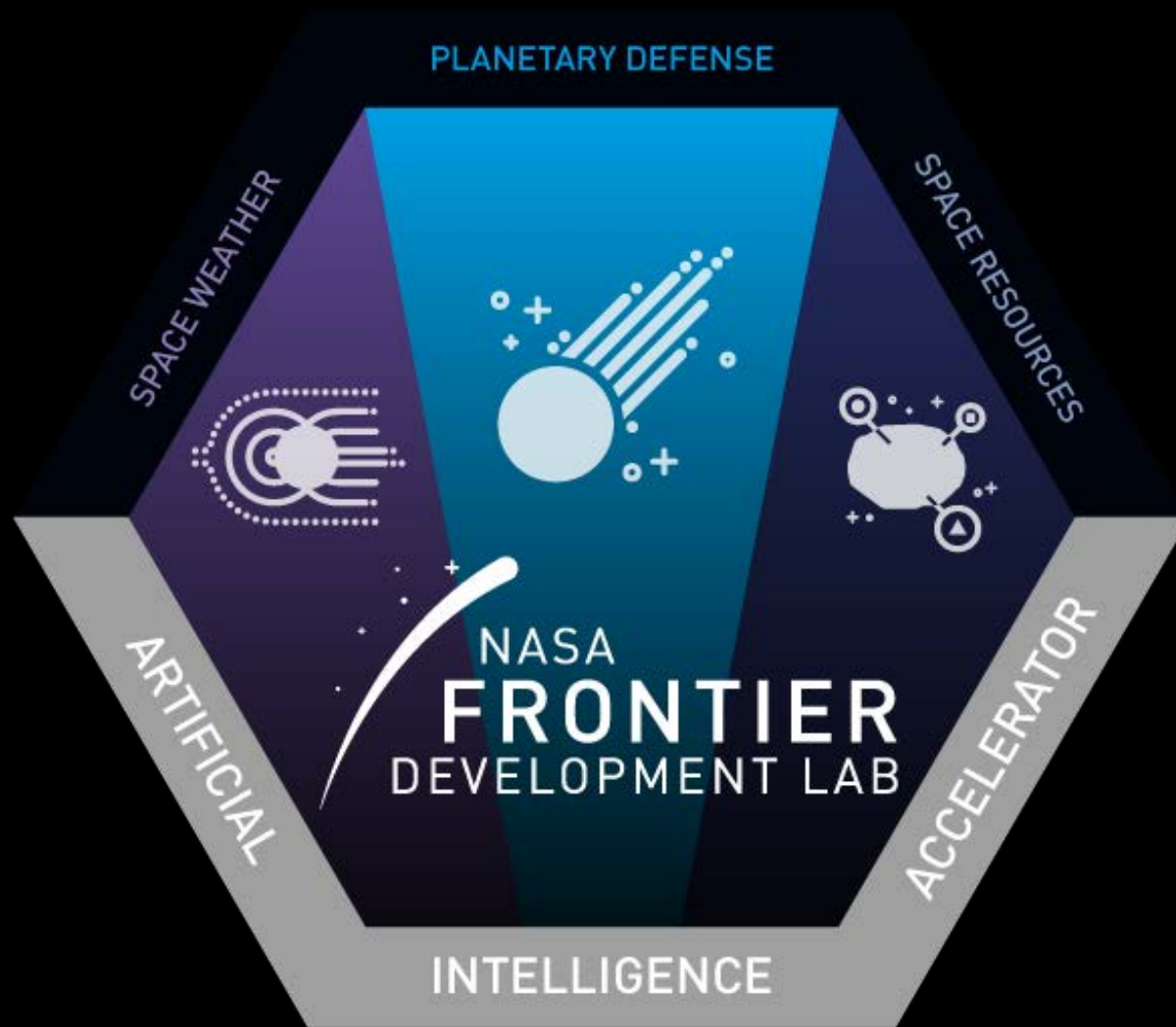


Nintendo announces Animal Crossing: Pocket Camp for smartphones

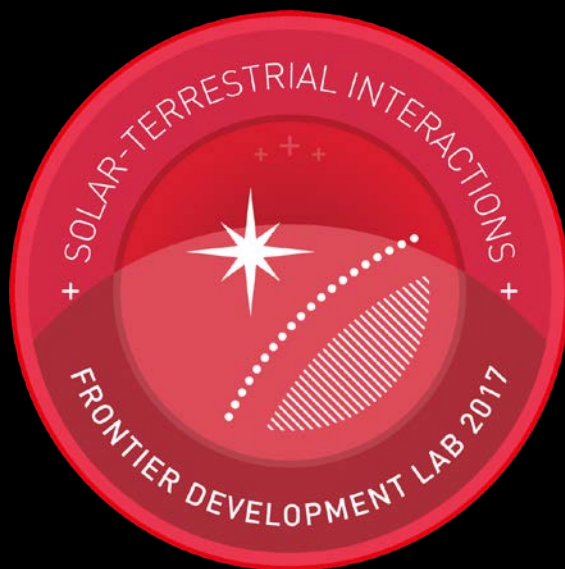
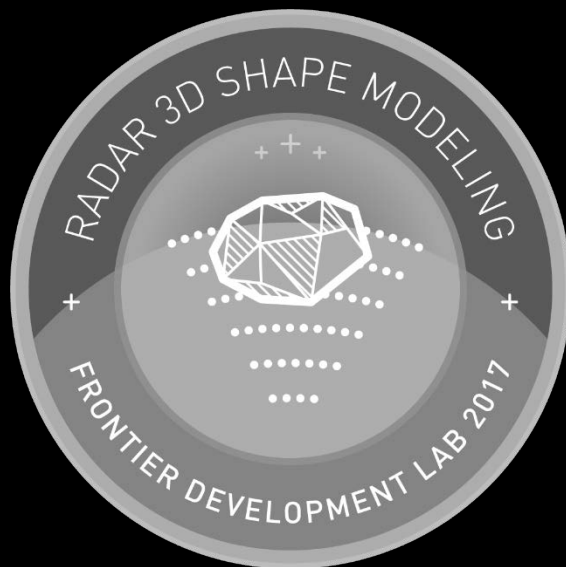
Deep learning has limitations too

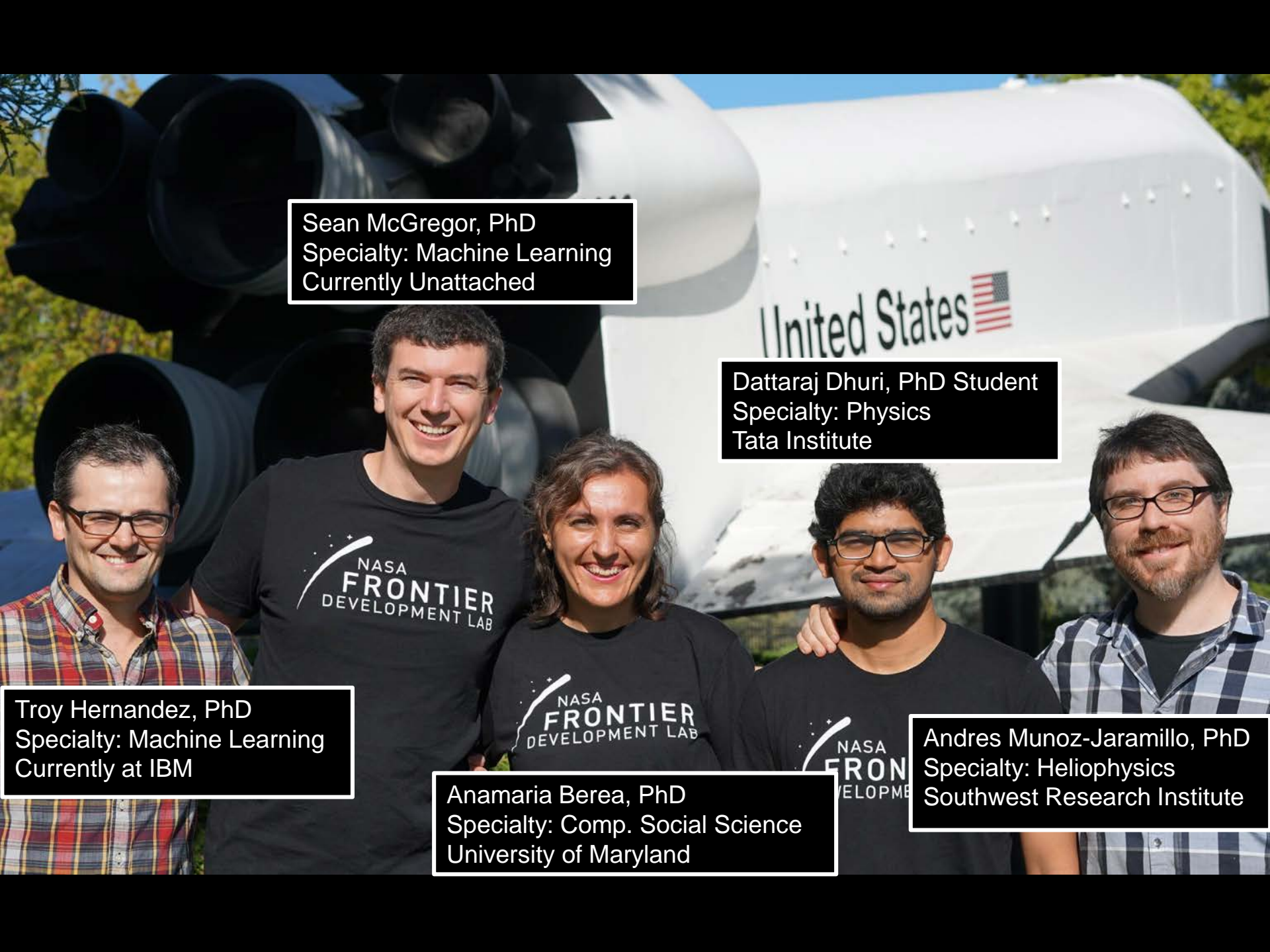
- Deep learning algorithms are completely naïve and single-minded in the way they learn.
- Training data selection is absolutely critical for their success.

PROOF OF CONCEPT APPLICATION OF DEEP LEARNING TO SPACE WEATHER FORECAST



AN APPLIED RESEARCH ACCELERATOR DESIGNED TO ENHANCE NASA'S CAPABILITIES BY COMBINING THE EXPERTISE OF NASA, ACADEMIA, AND THE PRIVATE RESEARCH COMMUNITY.





Sean McGregor, PhD
Specialty: Machine Learning
Currently Unattached

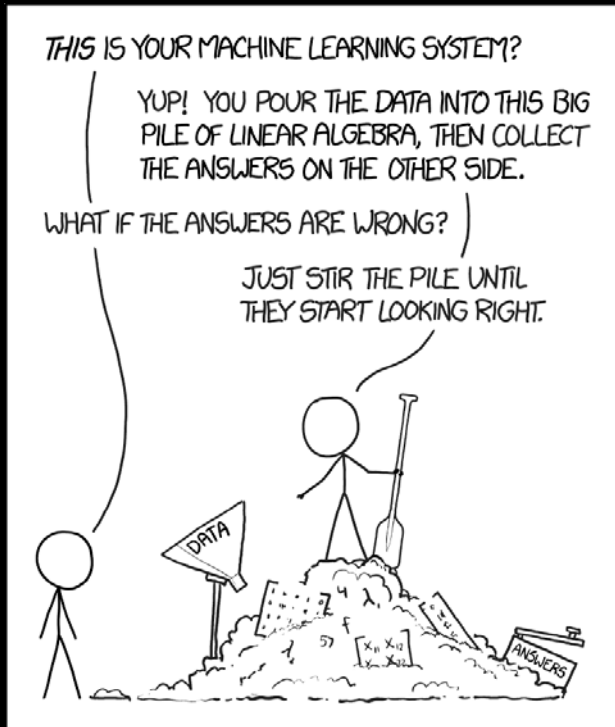
Dattaraj Dhuri, PhD Student
Specialty: Physics
Tata Institute

Troy Hernandez, PhD
Specialty: Machine Learning
Currently at IBM

Anamaria Berea, PhD
Specialty: Comp. Social Science
University of Maryland

Andres Munoz-Jaramillo, PhD
Specialty: Heliophysics
Southwest Research Institute

Interdisciplinary research is fun!



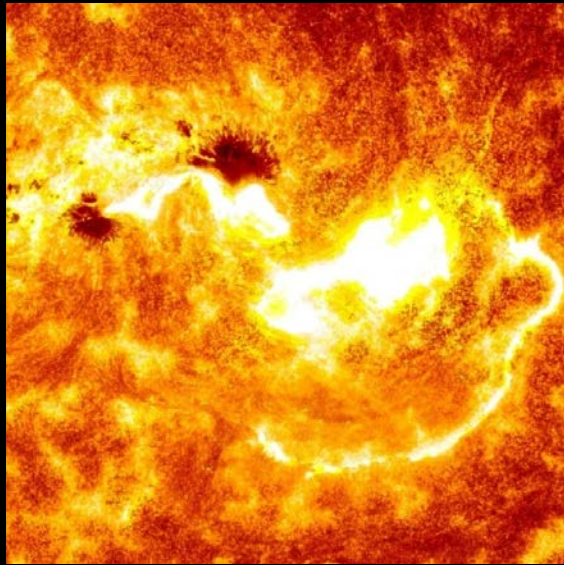
Heliophysicist's view of ML



Data scientist's view of HP

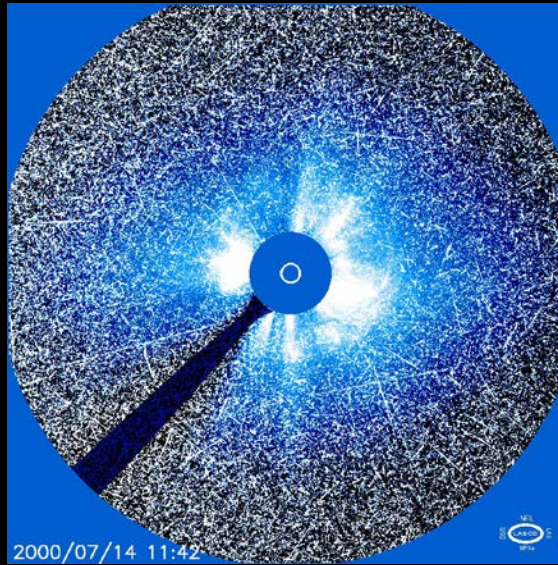
Which aspect of space weather to tackle?

FLARES



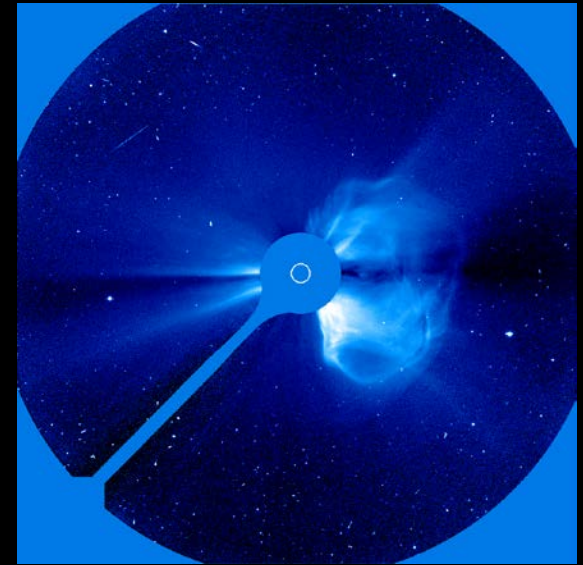
Speed of Light
No warning

ENERGETIC PARTICLES



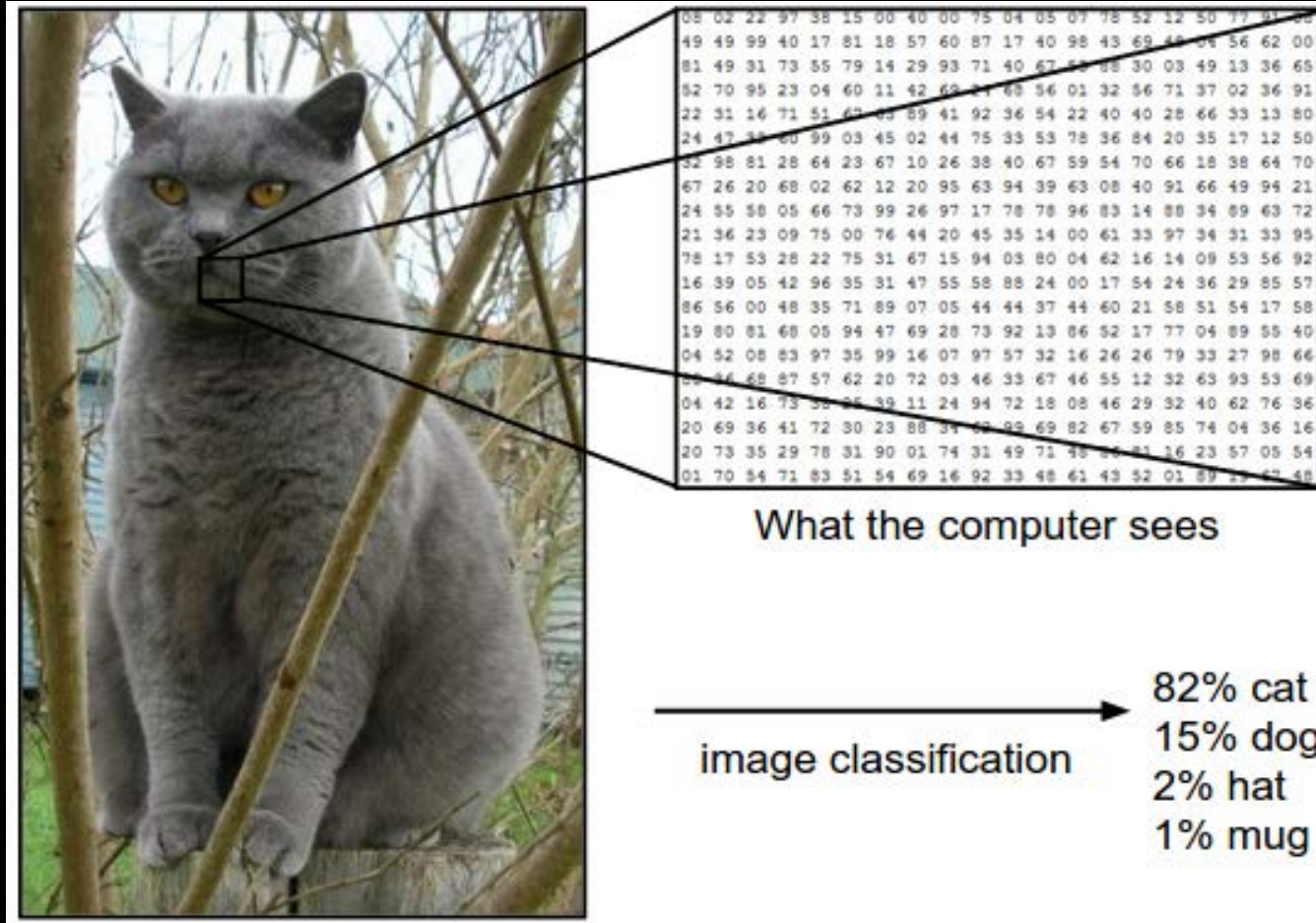
Relativistic speeds
20 minute warning

MASS EJECTIONS

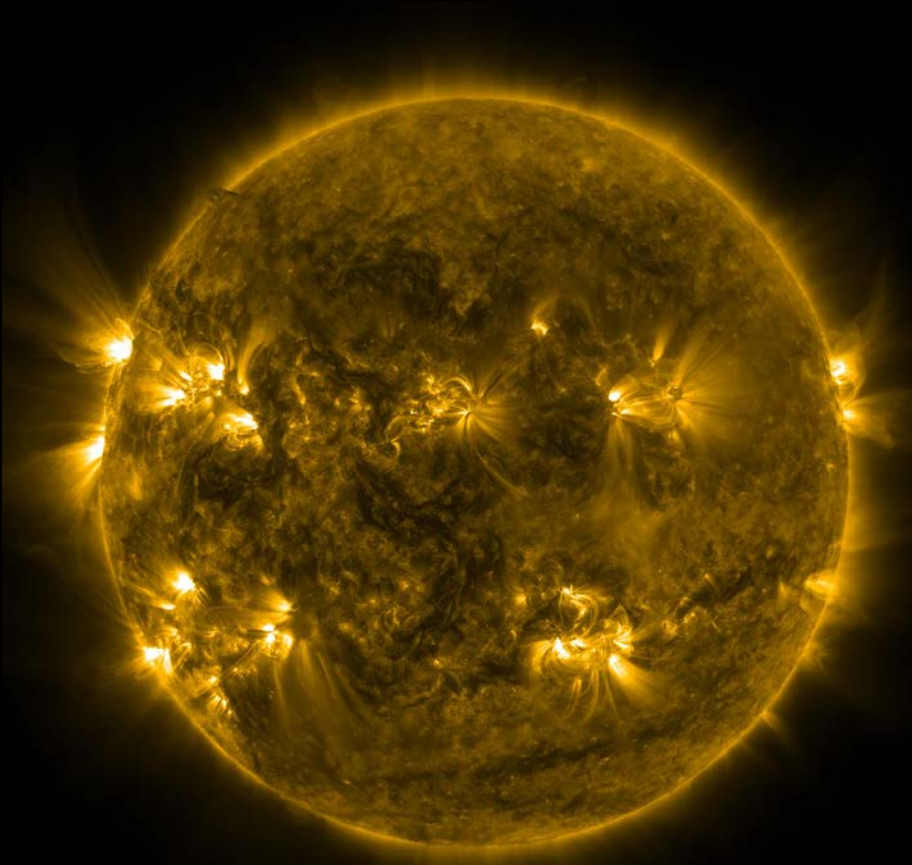


Non-relativistic speeds
20 hour warning

Deep learning and image data



Deep learning and image data



08	02	22	97	38	15	00	40	00	75	04	05	07	78	52	12	50	77	01	88
49	49	99	40	17	81	18	57	60	87	17	40	98	43	69	42	04	56	62	00
81	49	31	73	55	79	14	29	93	71	40	67	55	88	30	03	49	13	36	65
52	70	95	23	04	60	11	42	68	24	68	56	01	32	56	71	37	02	36	91
22	31	16	71	51	65	05	89	41	92	36	54	22	40	40	28	66	33	13	80
24	47	33	60	99	03	45	02	44	75	33	53	78	36	84	20	35	17	12	50
32	98	81	28	64	23	67	10	26	38	40	67	59	54	70	66	18	38	64	70
67	26	20	68	02	62	12	20	95	63	94	39	63	08	40	91	66	49	94	21
24	55	58	05	66	73	99	26	97	17	78	78	96	83	14	88	34	89	63	72
21	36	23	09	75	00	76	44	20	45	35	14	00	61	33	97	34	31	33	95
78	17	53	28	22	75	31	67	15	94	03	80	04	62	16	14	09	53	56	92
16	39	05	42	96	35	31	47	55	58	88	24	00	17	54	24	36	29	85	57
86	56	00	48	35	71	89	07	05	44	44	37	44	60	21	58	51	54	17	58
19	80	81	68	05	94	47	69	28	73	92	13	86	52	17	77	04	89	55	40
04	52	08	83	97	35	99	16	07	97	57	32	16	26	26	79	33	27	98	66
29	16	68	87	57	62	20	72	03	46	33	67	46	55	12	32	63	93	53	69
04	42	16	73	51	05	39	11	24	94	72	18	08	46	29	32	40	62	76	36
20	69	36	41	72	30	23	88	34	80	89	69	82	67	59	85	74	04	36	16
20	73	35	29	78	31	90	01	74	31	49	71	48	16	81	16	23	57	05	54
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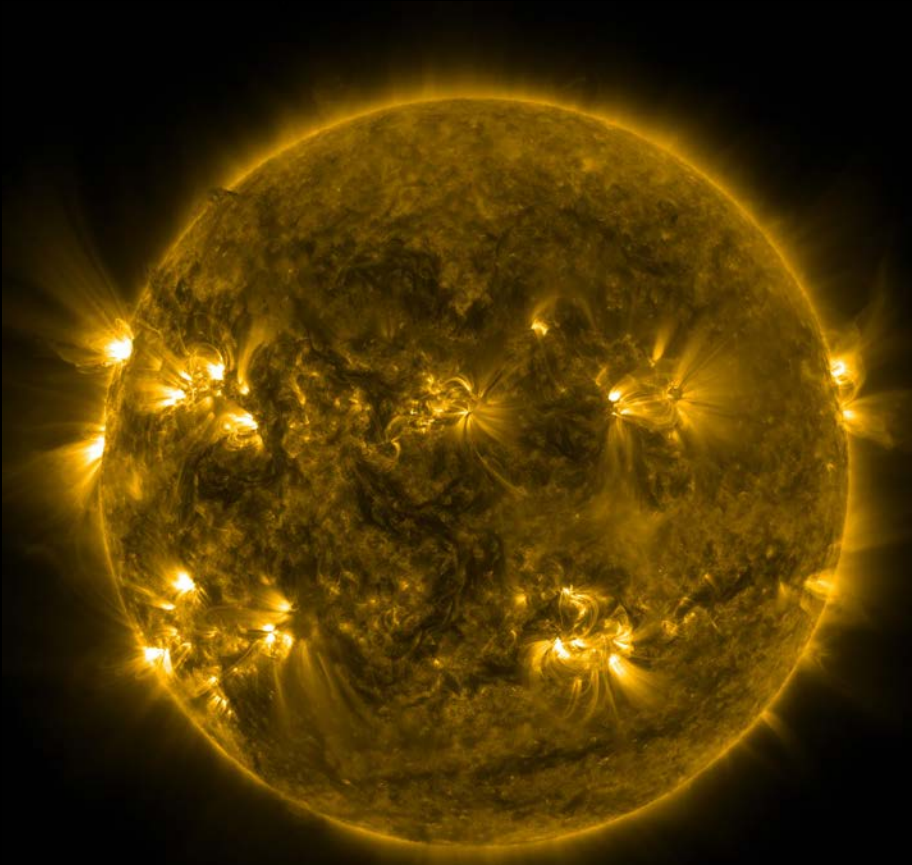
What the computer sees

image classification

82% cat
15% dog
2% hat
1% mug

SDO/AIA 171 2012-11-13 16:30:12 UT

Deep learning and image data



What the computer sees

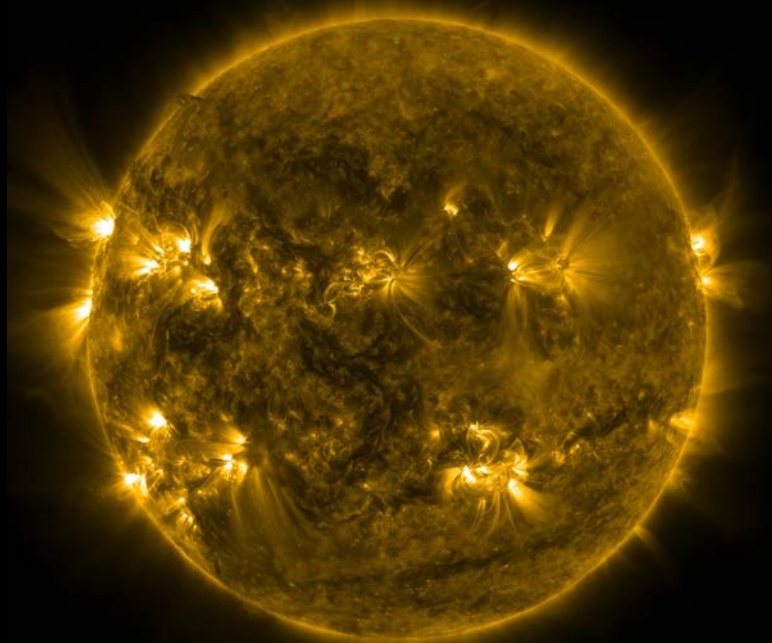
X-Ray Flux

3.2×10^{-5}
 W m^{-2}

SDO/AIA 171 2012-11-13 16:30:12 UT

Convolutional Neural Networks

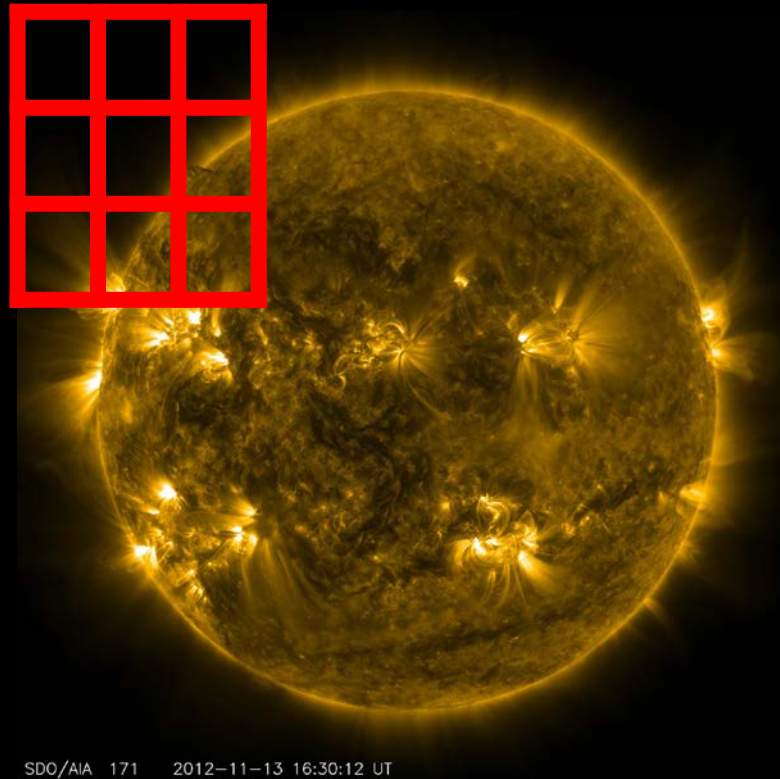
Neural networks with layers made of tunable convolution filters



SDO/AIA 171 2012-11-13 16:30:12 UT

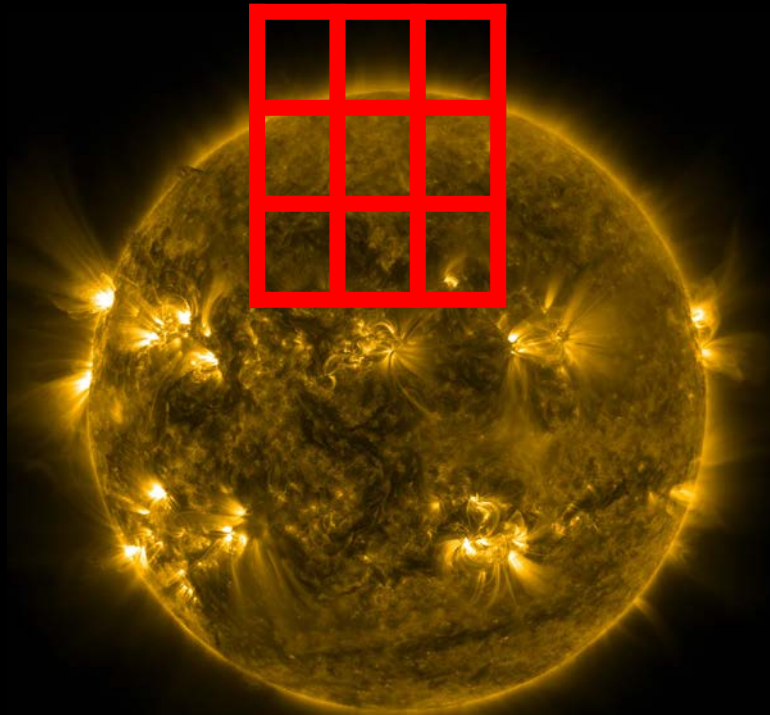
Convolutional Neural Networks

Neural networks with layers made of tunable convolution filters



Convolutional Neural Networks

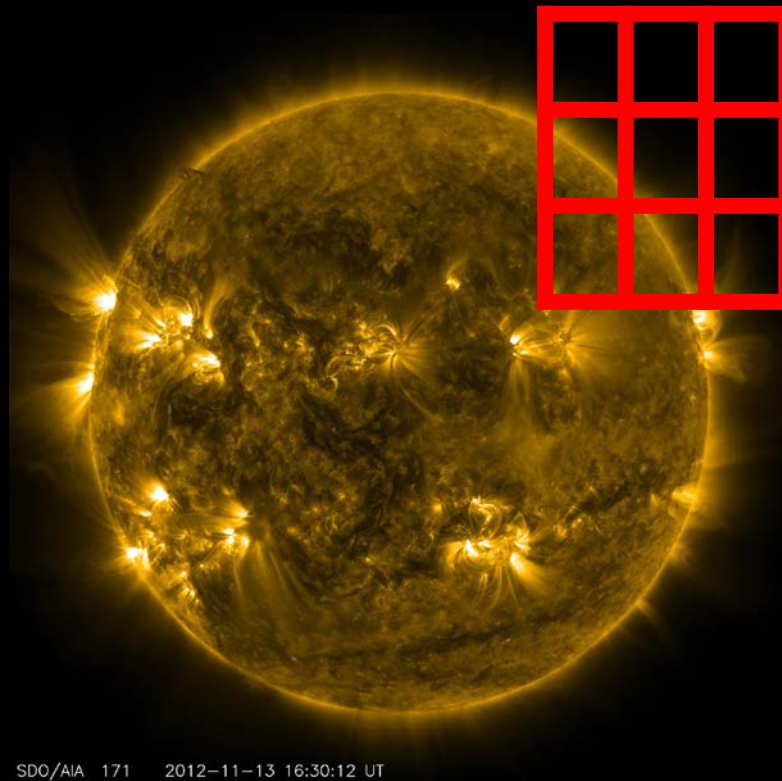
Neural networks with layers made of tunable convolution filters



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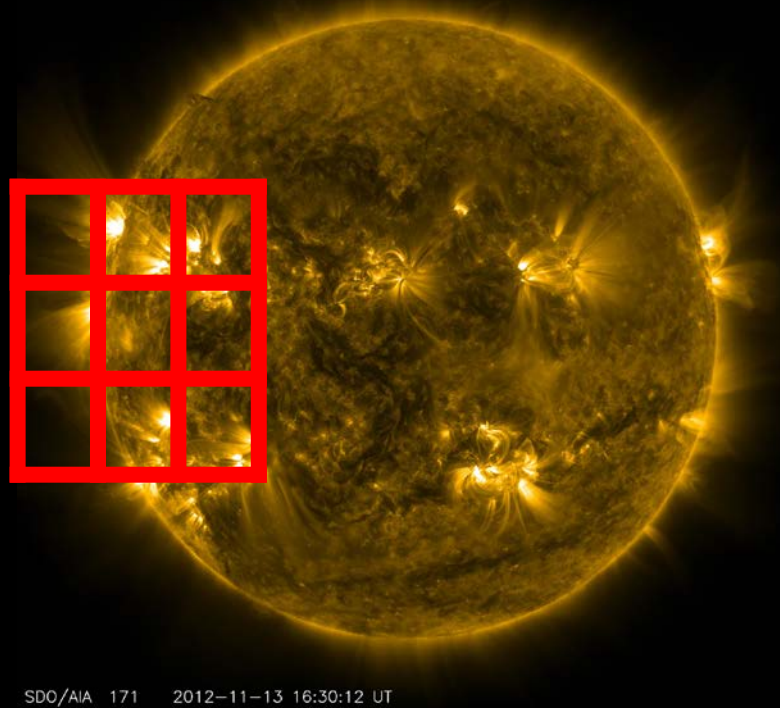
Convolutional Neural Networks

Neural networks with layers made of tunable convolution filters



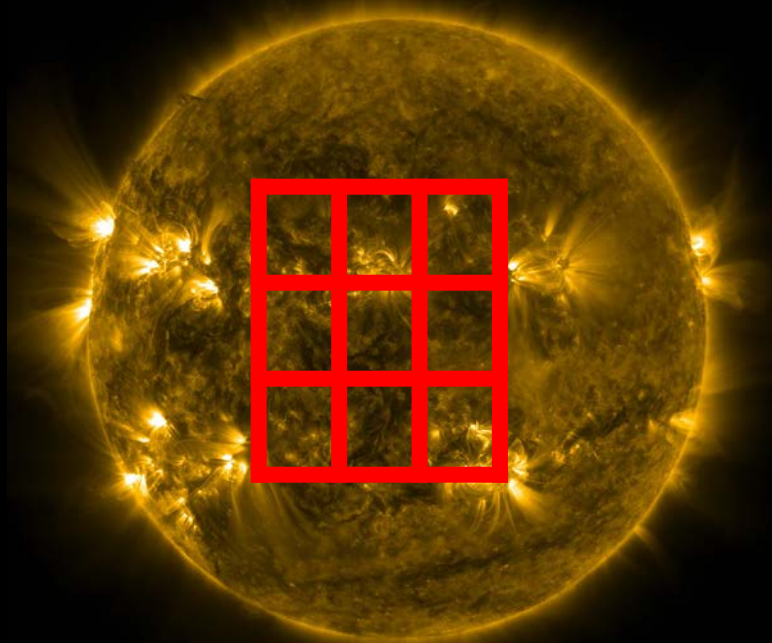
Convolutional Neural Networks

Neural networks with layers made of tunable convolution filters



Convolutional Neural Networks

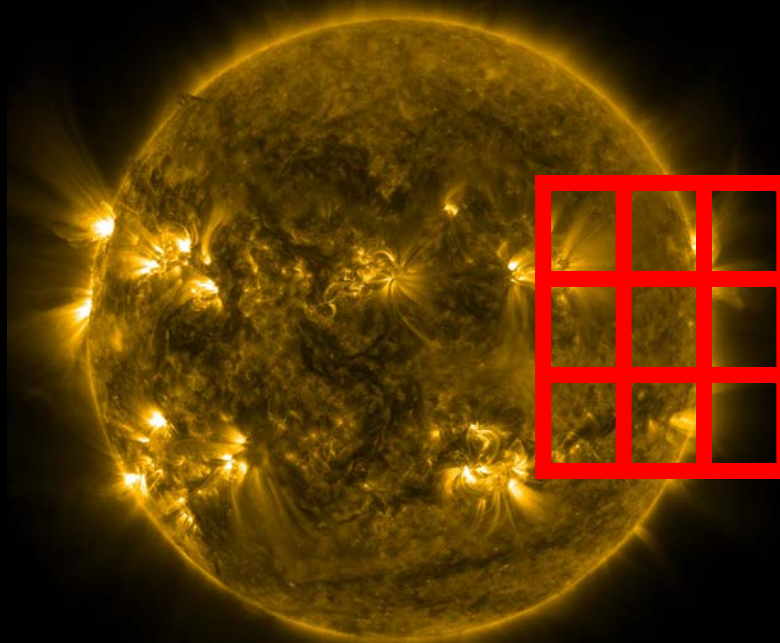
Neural networks with layers made of tunable convolution filters



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Convolutional Neural Networks

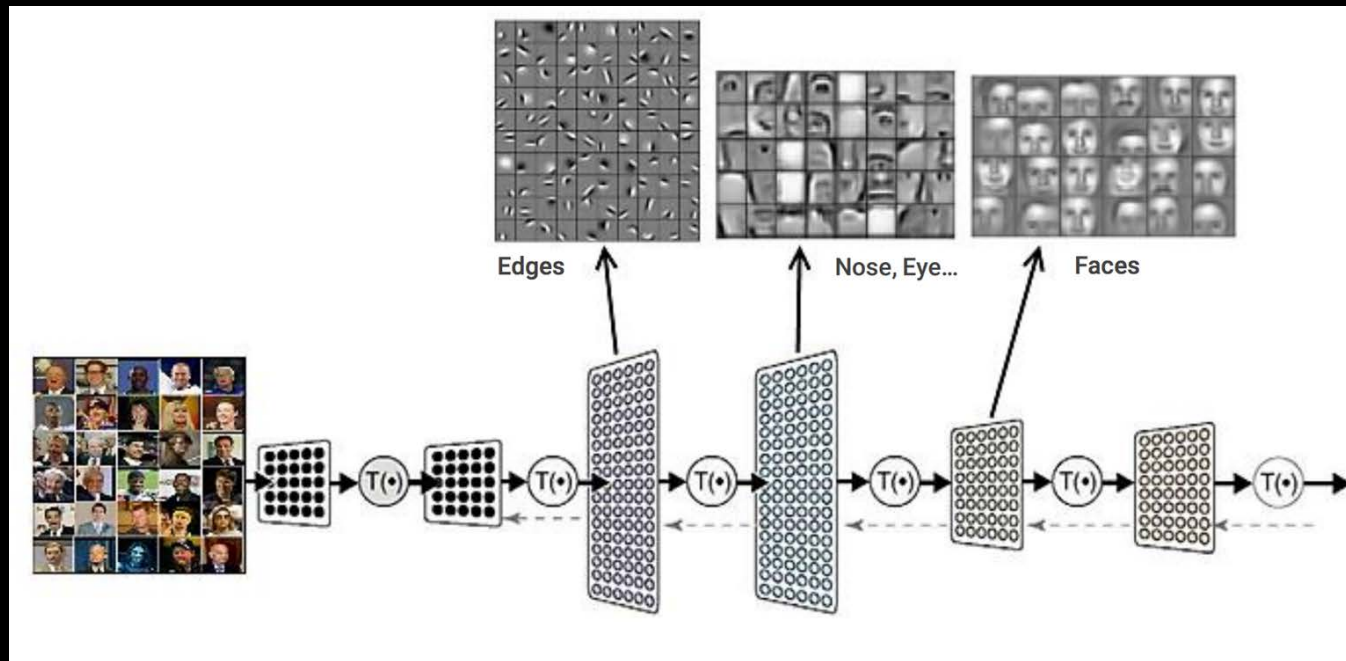
Neural networks with layers made of tunable convolution filters



SDO/AIA 171 2012-11-13 16:30:12 UT

Convolutional Neural Networks

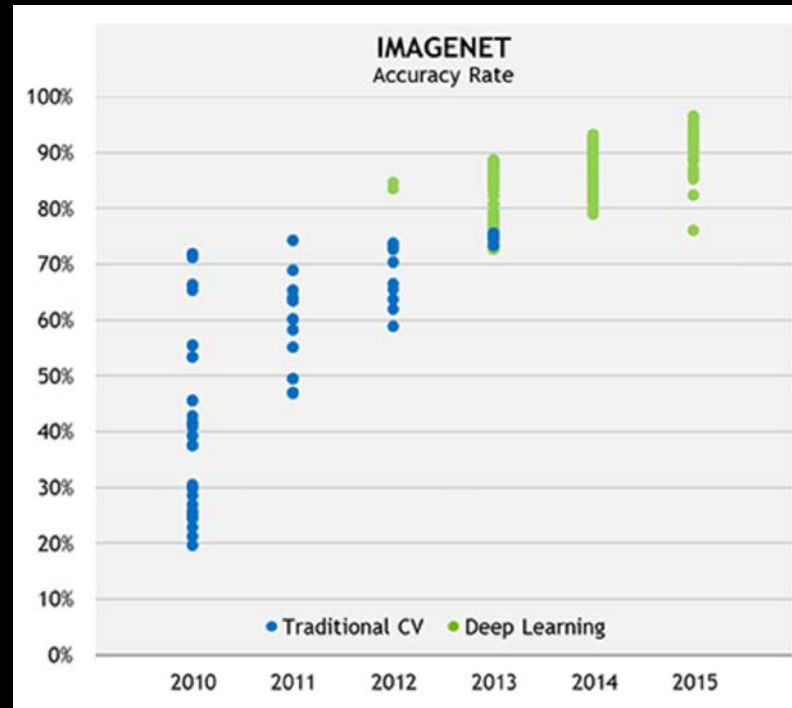
Neural networks with layers made of tunable convolution filters



Several convolutional layers allow the neural network to recognize features of increased complexity

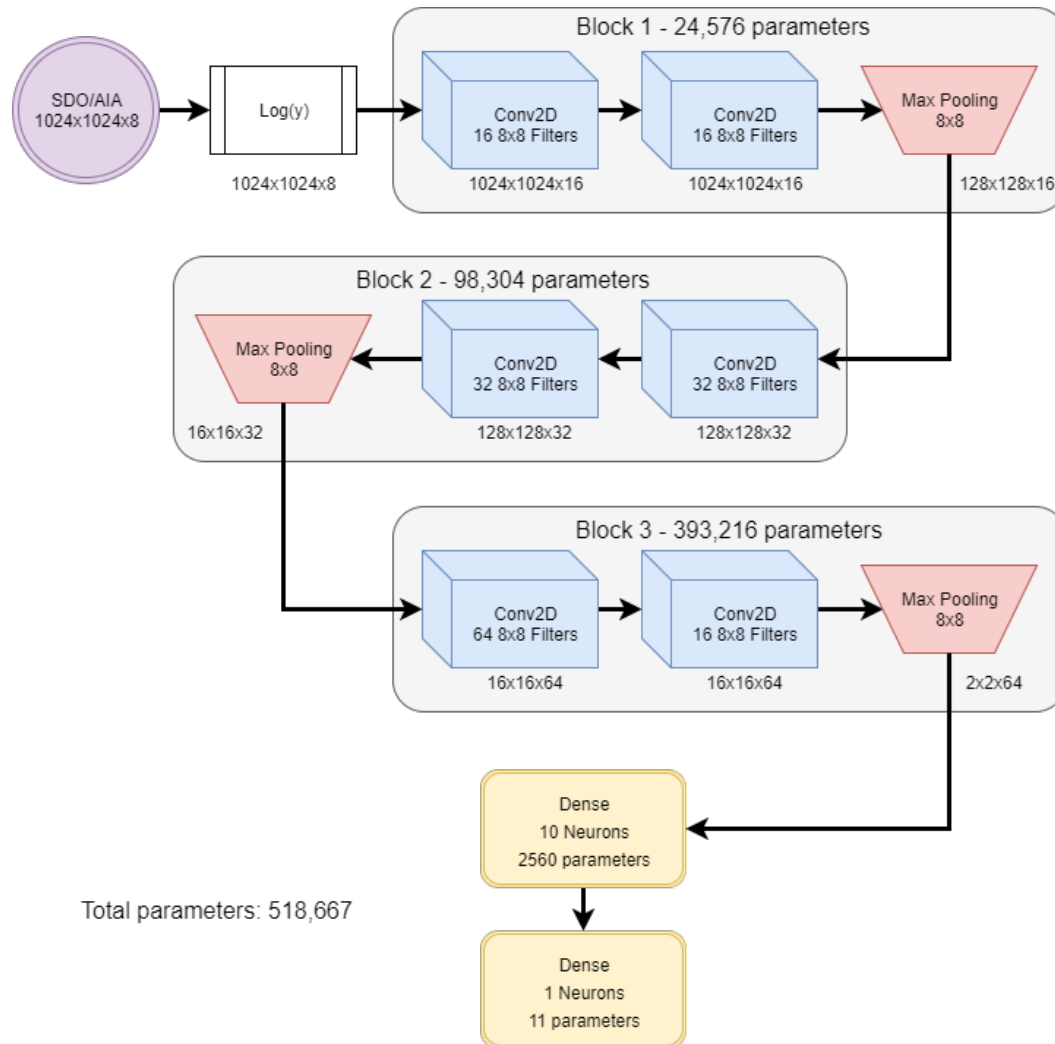
Convolutional Neural Networks

Neural networks with layers made of tunable convolution filters



CNNs have revolutionized the way we do image classification.

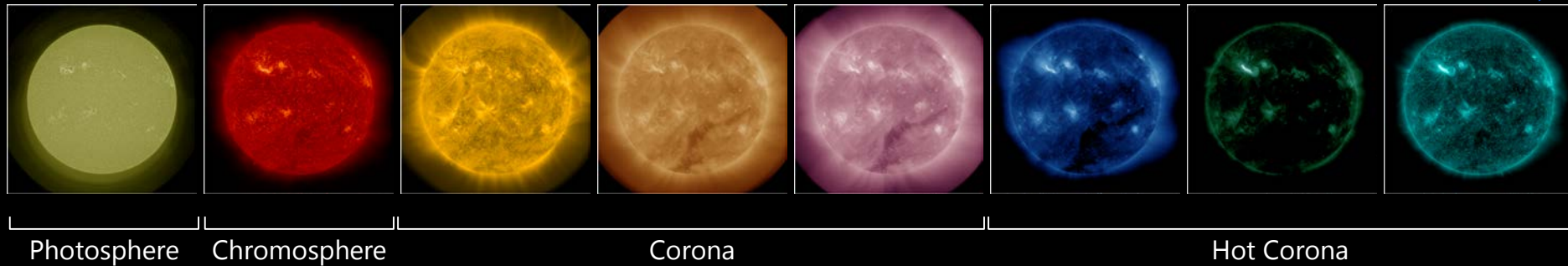
FlareNet



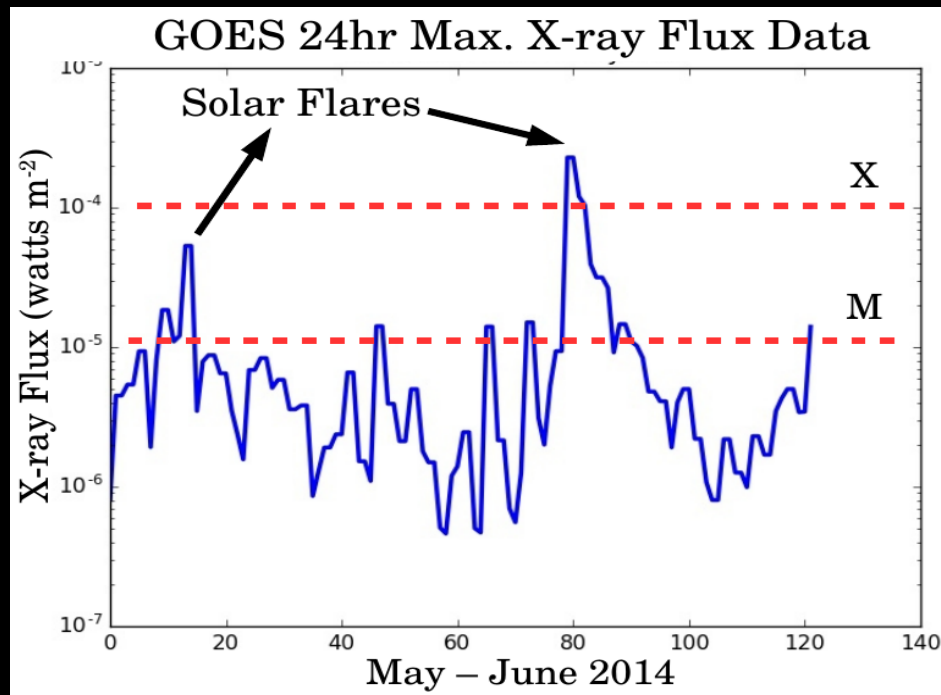
Input and output data

Input data: SDO/AIA

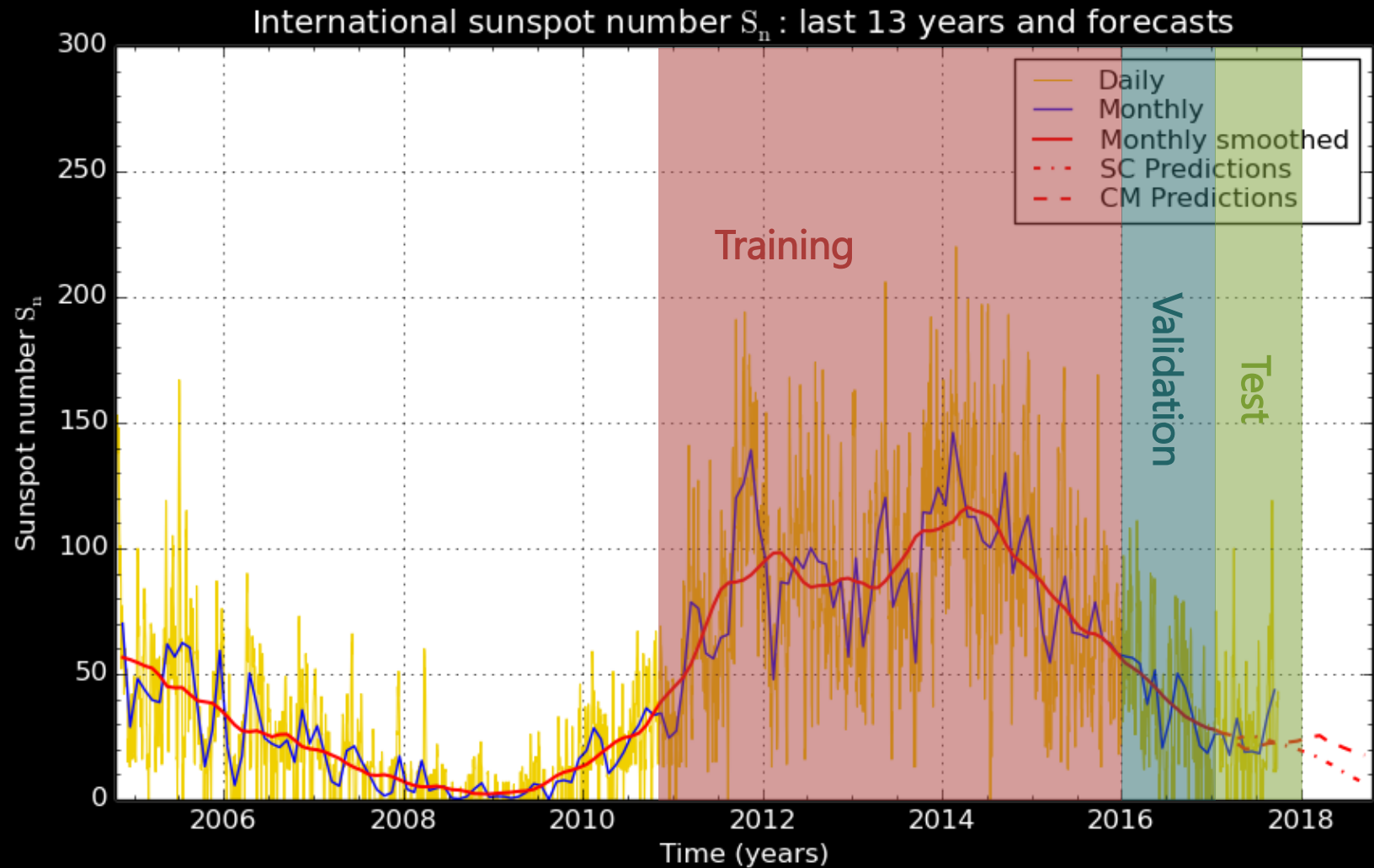
Hotter and Higher



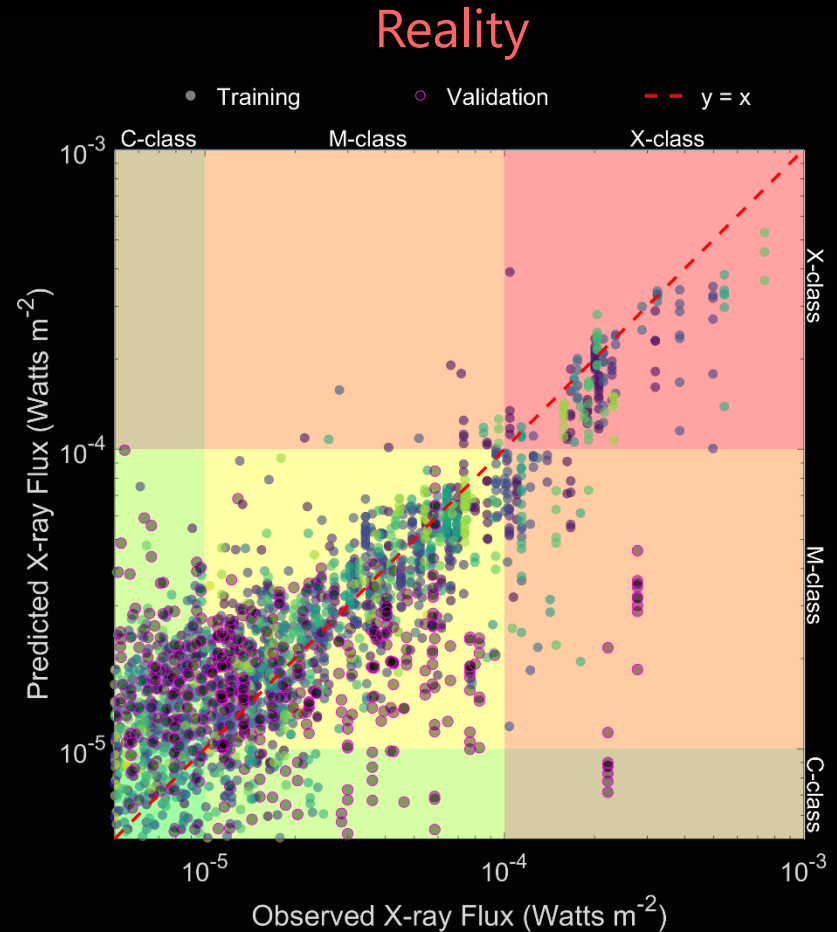
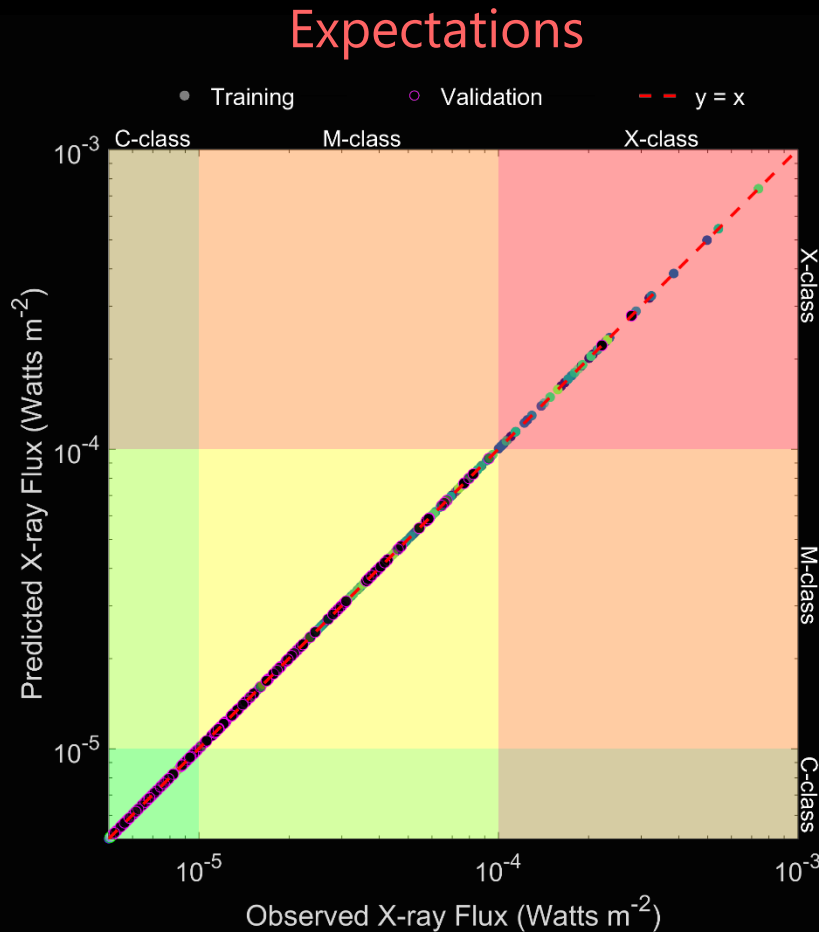
Output data: Goes
X-Ray flux



Training, Validation and Test

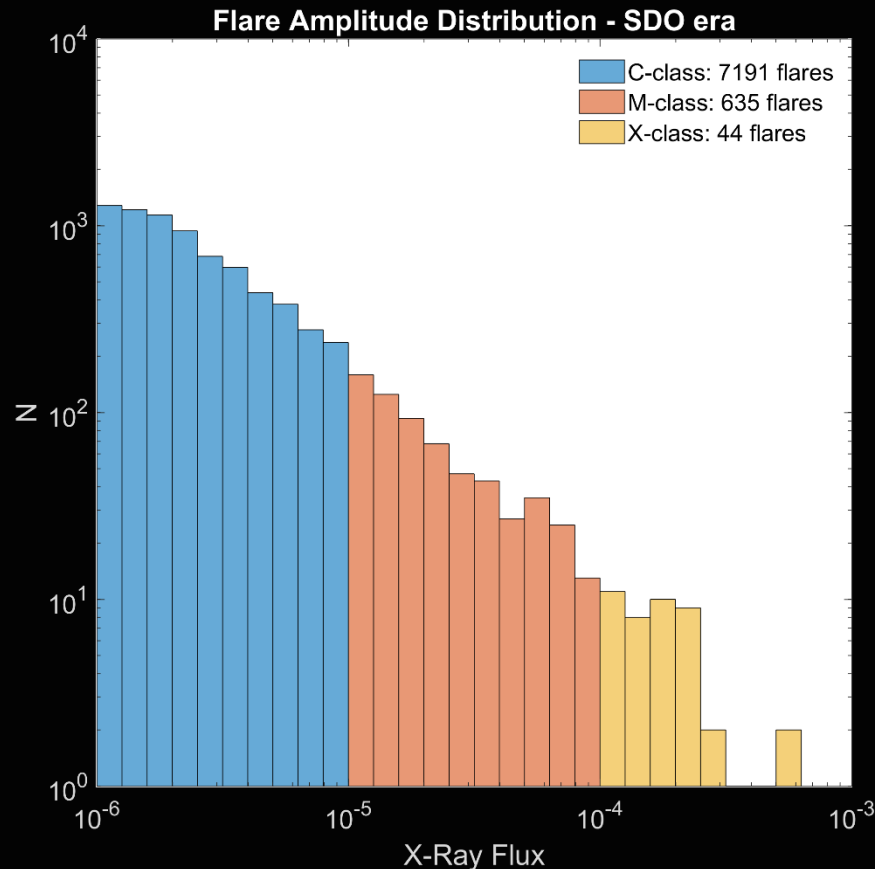


Results: Expectations vs. Reality



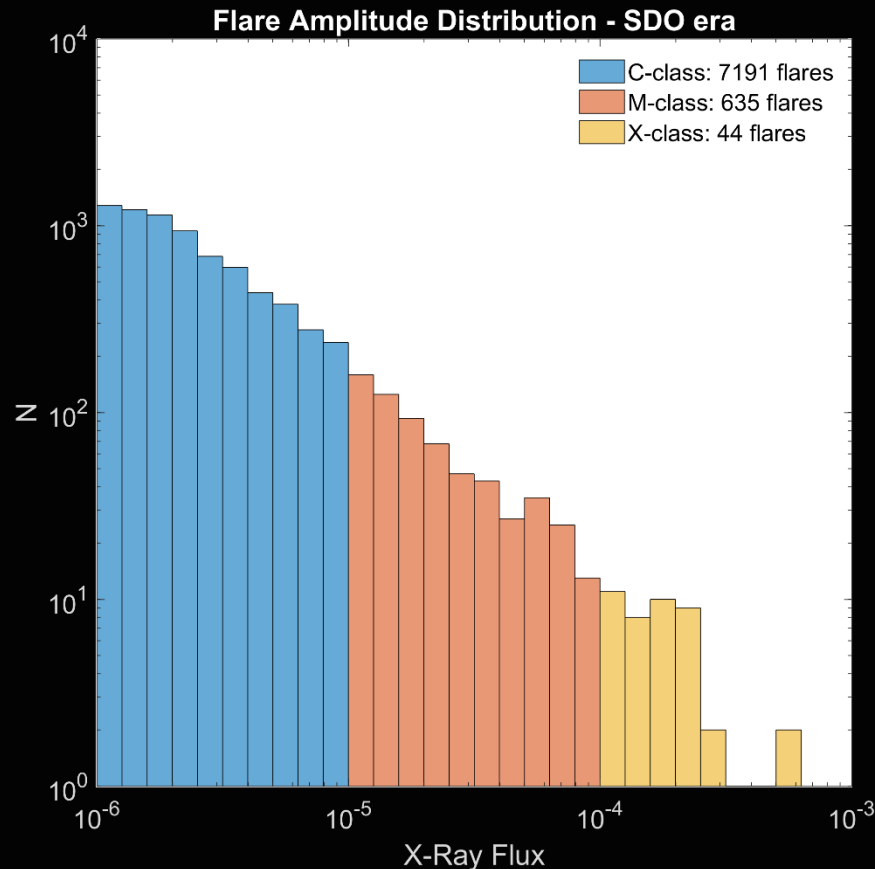
We are achieving some generalization, but finding challenging to make a clean connection between images and flares due to class imbalance

Results: Expectations vs. Reality



We are achieving some generalization, but finding challenging to make a clean connection between images and flares due to class imbalance

Results: Expectations vs. Reality



Data shouldn't be only big,
it should also be long!

Grad-CAM: Why did you say that?

Visual Explanations from Deep Networks via Gradient-based Localization

Ramprasaath R. Selvaraju

Abhishek Das

Ramakrishna Vedantam

Michael Cogswell

Devi Parikh

Dhruv Batra

Virginia Tech

{ram21, abhshkdz, vrama91, cogswell, parikh, dbatra}@vt.edu

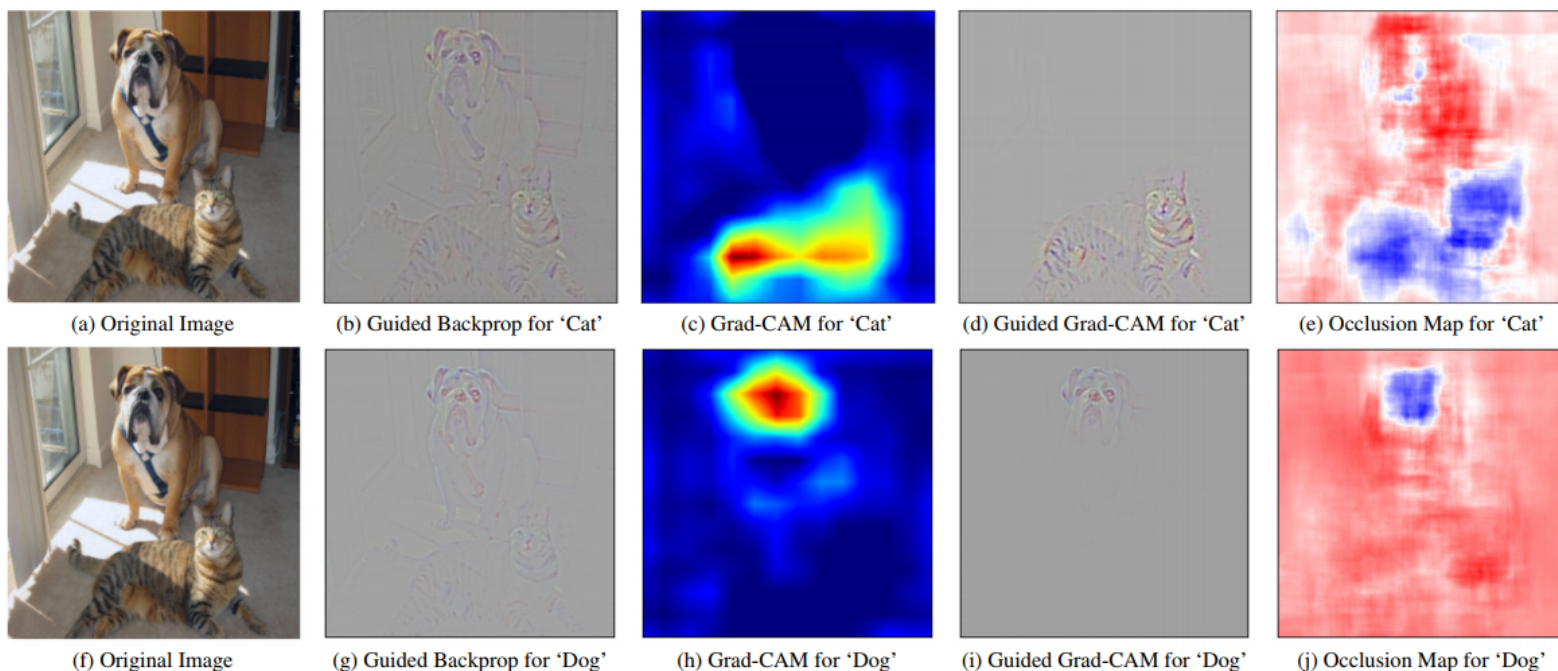
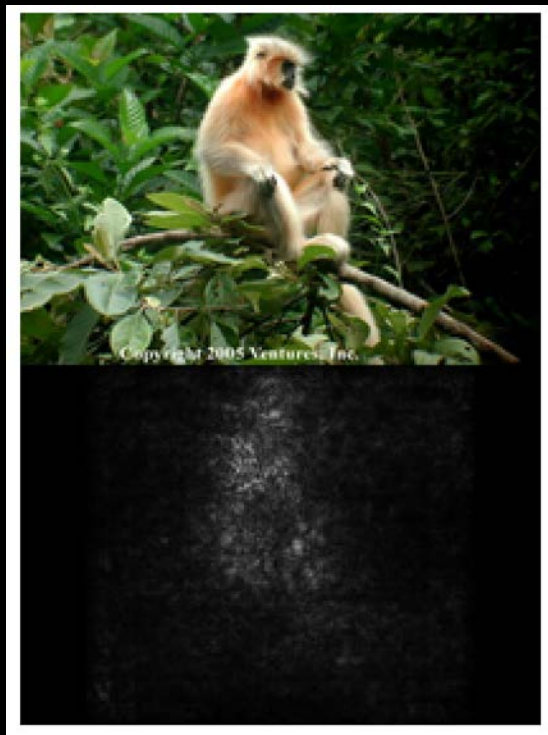


Figure 1. (a) Original image with a cat and a dog. (b-e) Support for the cat category according to various visualizations. (b) Guided Backpropagation [38]: provides high-resolution visualization of contributing features, (c) Grad-CAM (Ours): localizes class-discriminative regions, (d) Combining (b) and (c) gives Guided Grad-CAM, which gives high-resolution visualizations that are class-discriminative. Interestingly, the localizations achieved by our Grad-CAM technique (c) are very similar to results from occlusion sensitivity (e), while being much cheaper to compute. Note that in (e), blue corresponds to evidence for the class while in (c) blue indicates regions with low score for the class. Figure best viewed in color.

MINING NEURAL NETWORKS

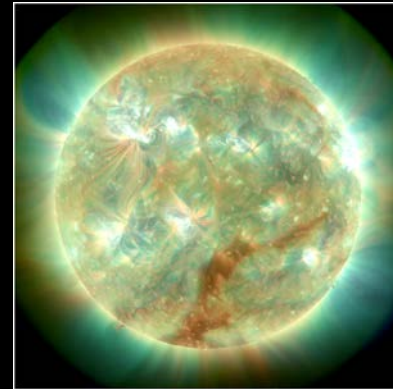
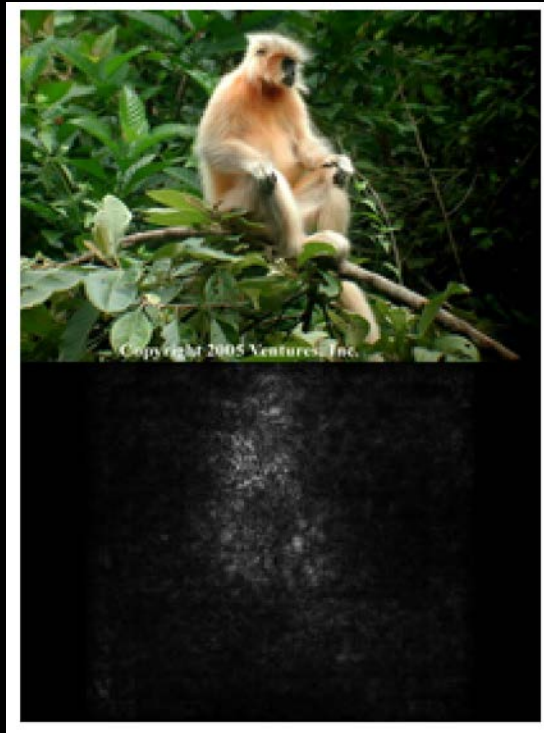
Mining Neural Networks: Saliency



What does a convolutional neural network pay attention to?

Simonyan, K., Vedaldi, A., & Zisserman, A. (2013). Deep Inside Convolutional Networks: Visualising Image Classification Models and Saliency Maps.

Mining Neural Networks: Saliency

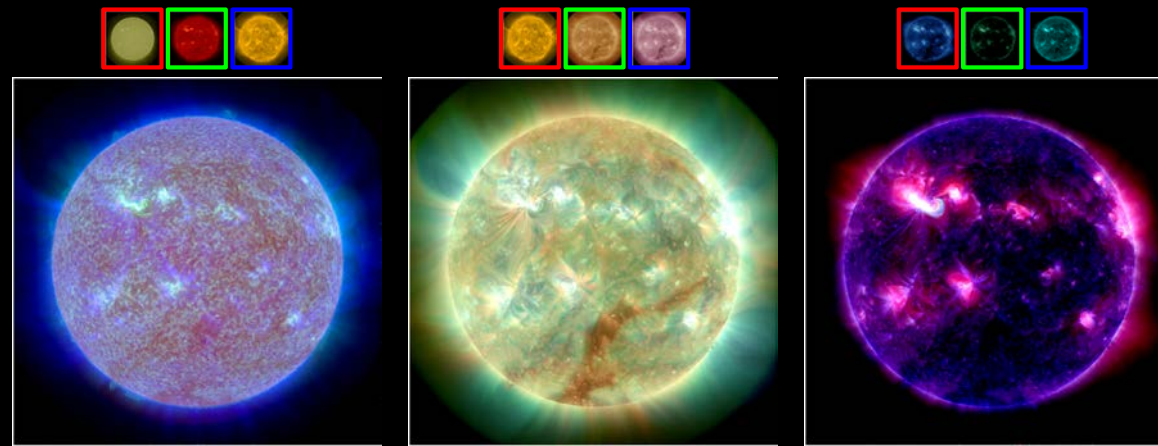
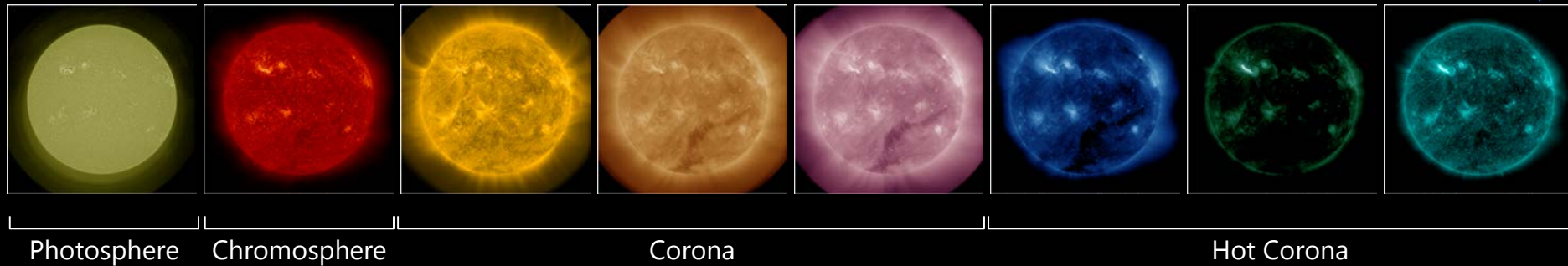


?

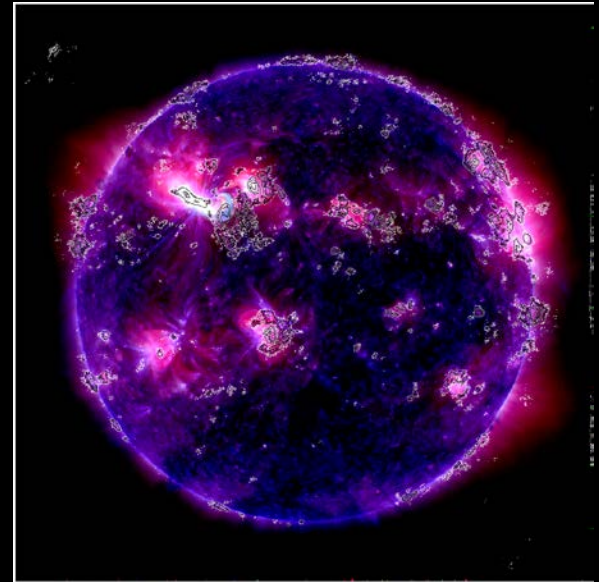
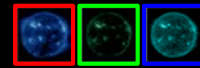
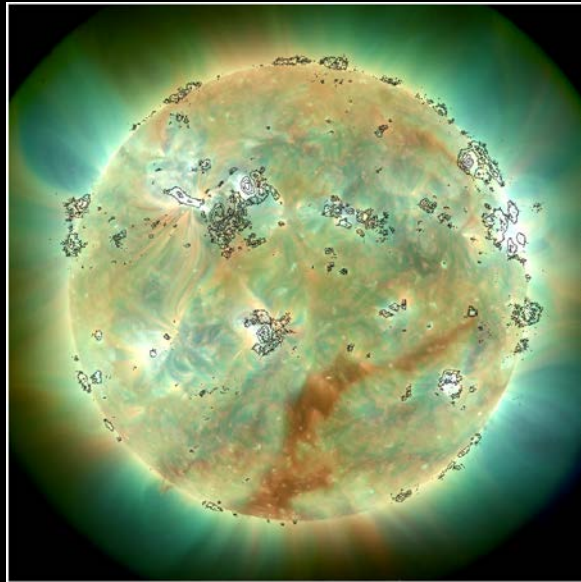
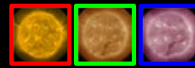
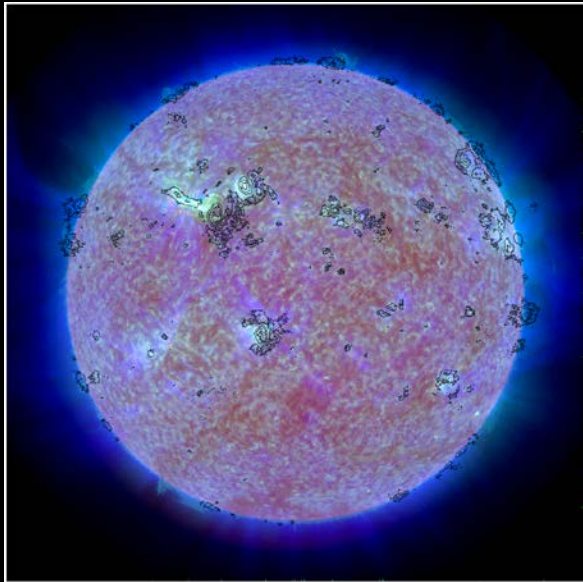
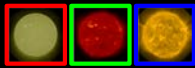
Simonyan, K., Vedaldi, A., & Zisserman, A. (2013). Deep Inside Convolutional Networks: Visualising Image Classification Models and Saliency Maps.

Mining Neural Networks: Saliency

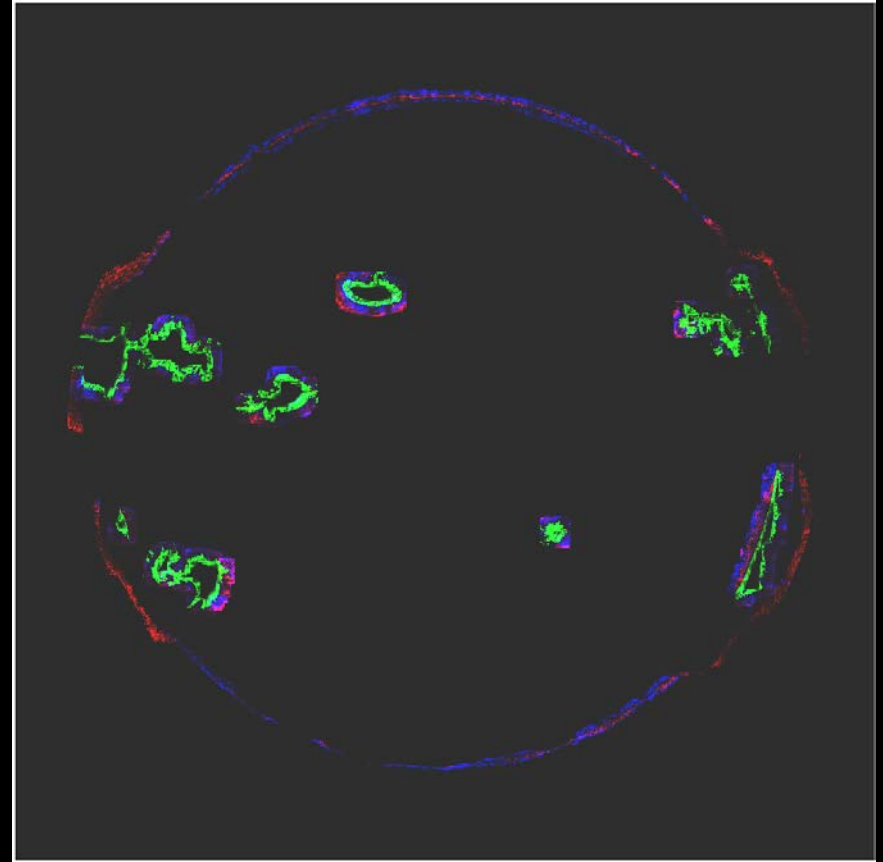
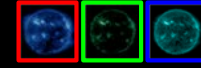
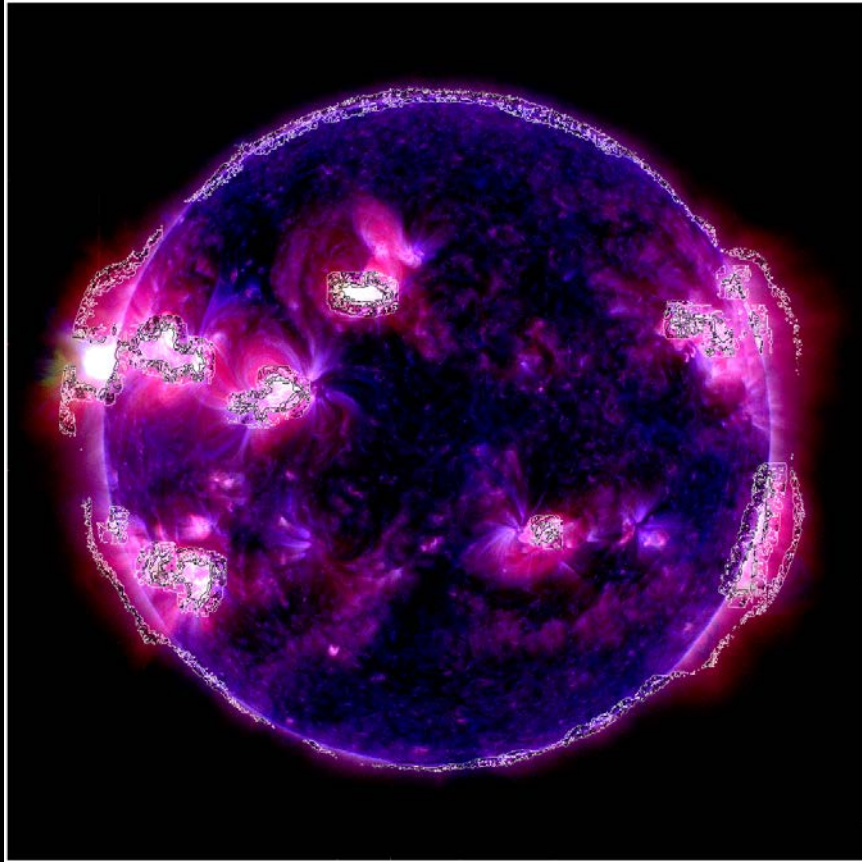
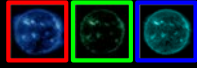
Hotter and Higher



Mining Neural Networks: Saliency

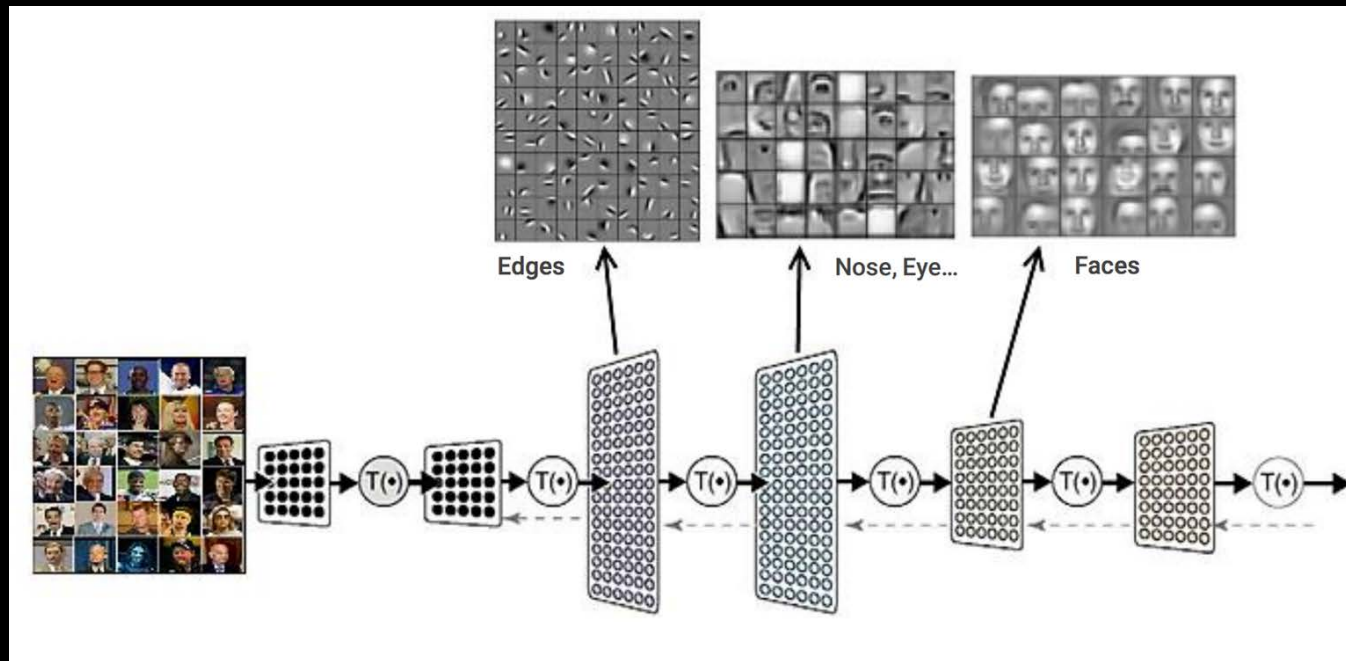


Mining Neural Networks: Saliency



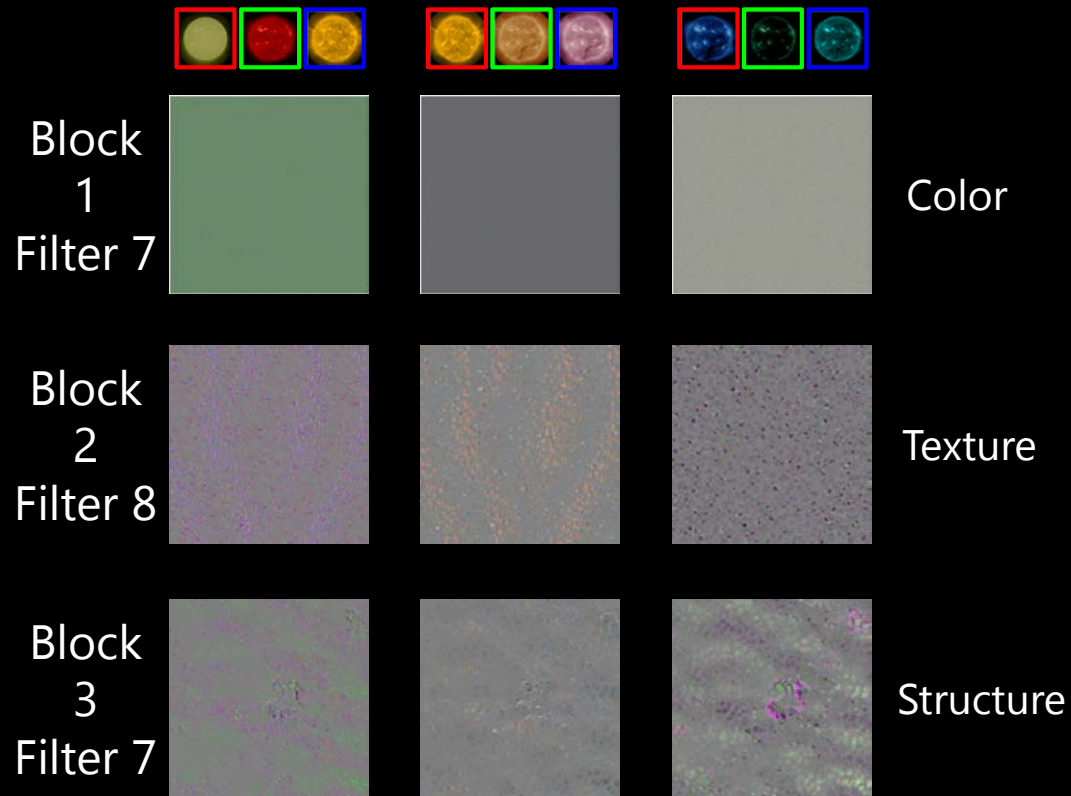
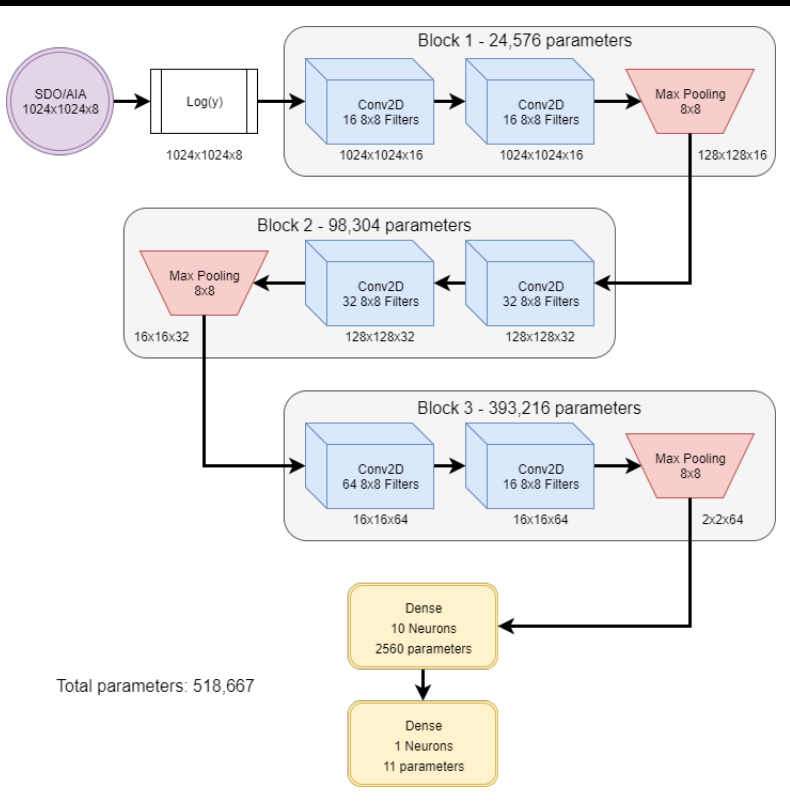
FlareNet is paying attention to the relative location of structures in different channels

Mining Neural Networks: Filter Activation

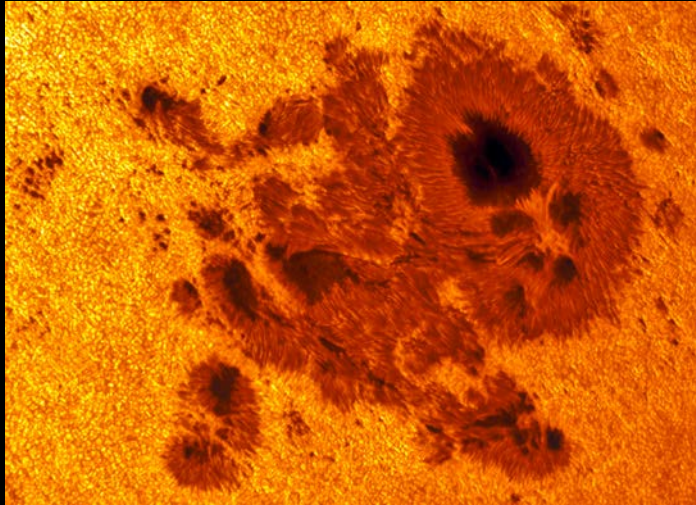


Several convolutional layers allow the neural network to recognize features of increased complexity

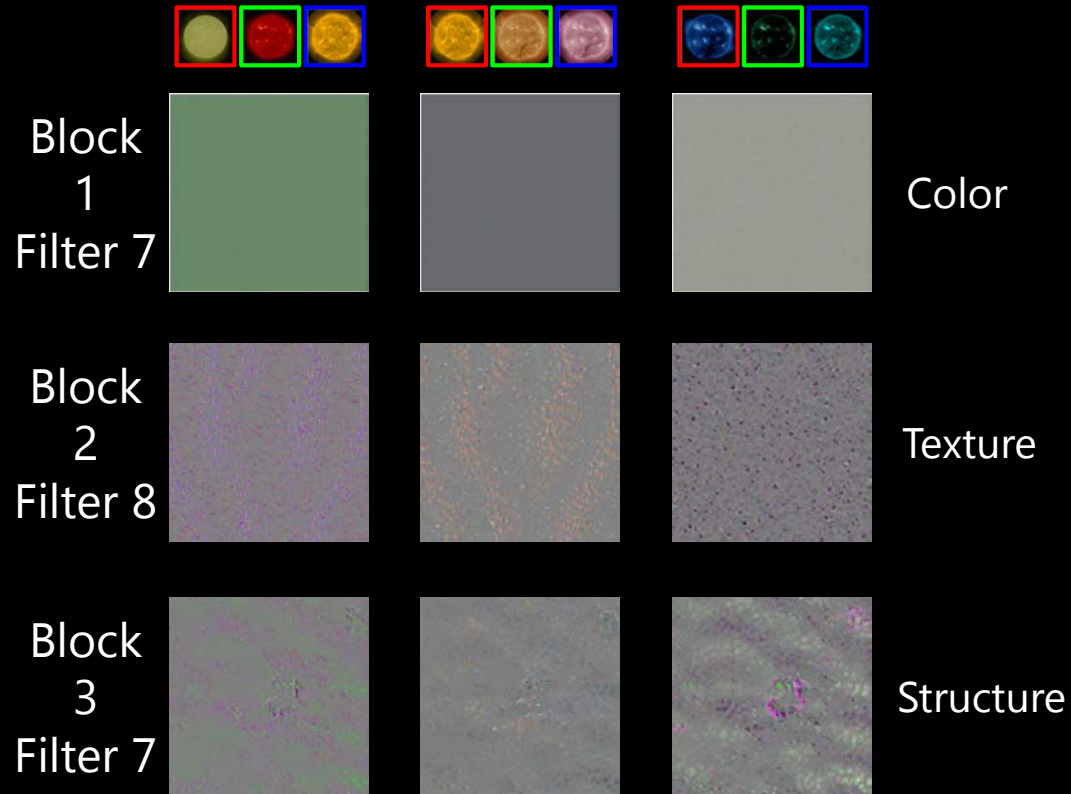
Mining Neural Networks: Filter Activation



Mining Neural Networks: Filter Activation



FlareNet learned the importance of active regions



What is deep learning?



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What it really is

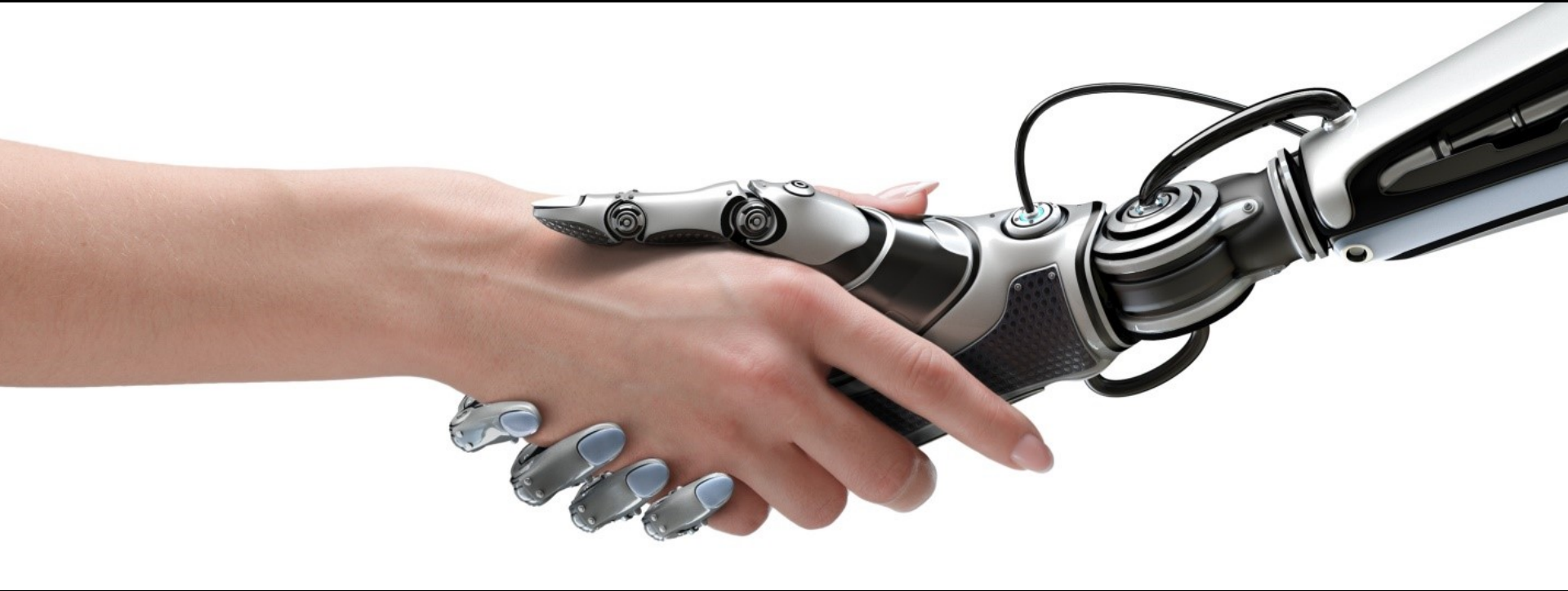


What skeptics think it is



What the public thinks it is

What is deep learning?



A revolutionary technique that will transform the way we interact with data (big or small)