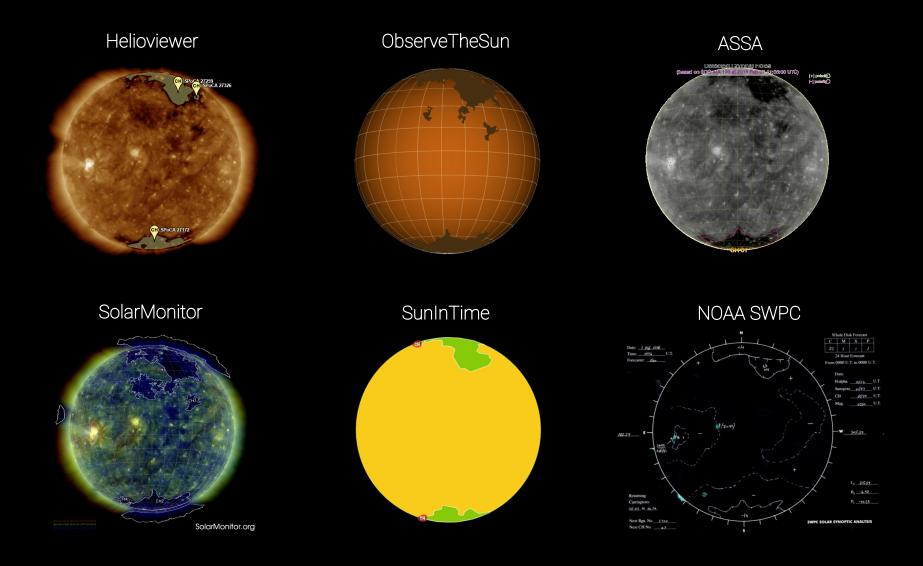


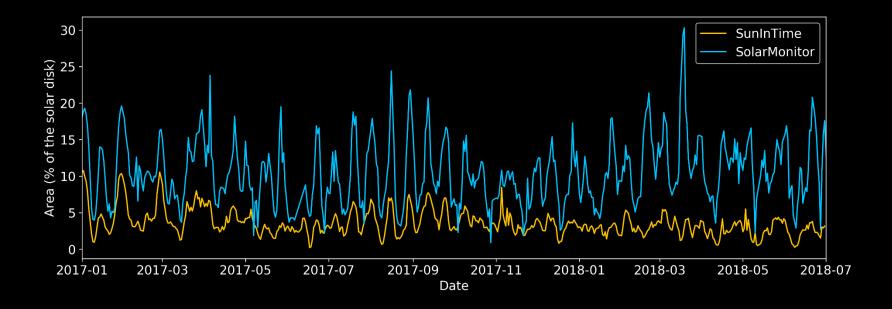
# Segmentation of solar disk images with a convolutional neural network

Egor ILLARIONOV (Moscow State University) Andrey TLATOV (Kislovodsk Mountain Solar Station)

## Examples of coronal holes segmentation for the same day

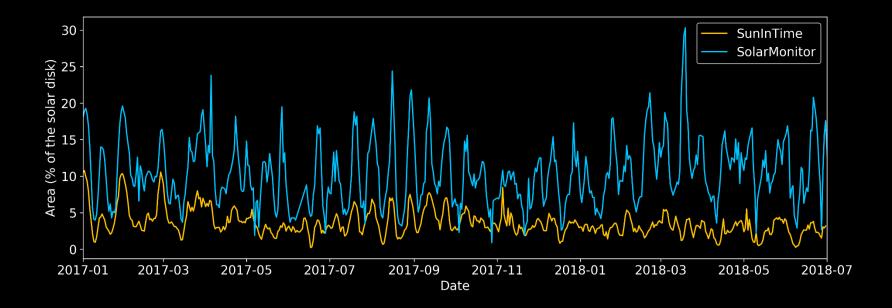


#### Total area of CHs according to SunInTime (Spoca) and SolarMonitor (CHIMERA)



Yet another algorithm?

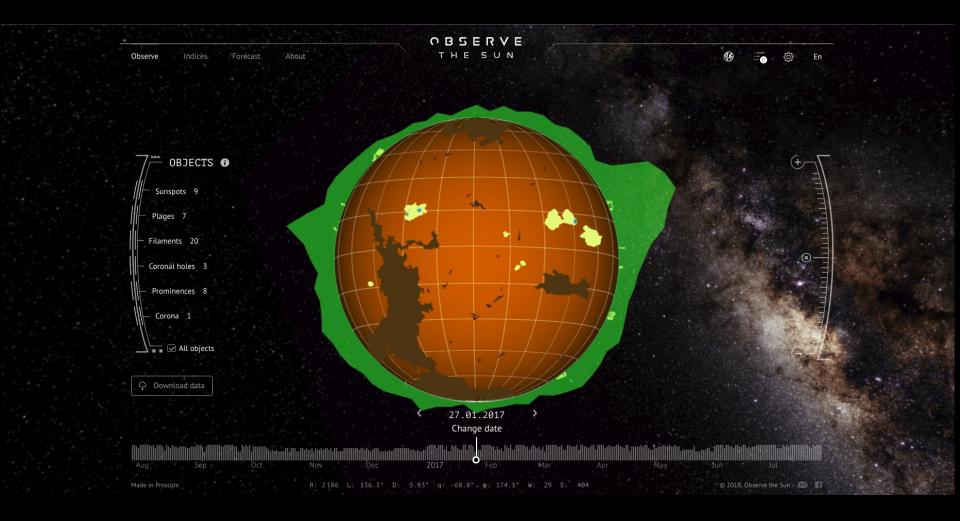
#### Total area of CHs according to SunInTime (Spoca) and SolarMonitor (CHIMERA)



Yet another algorithm?

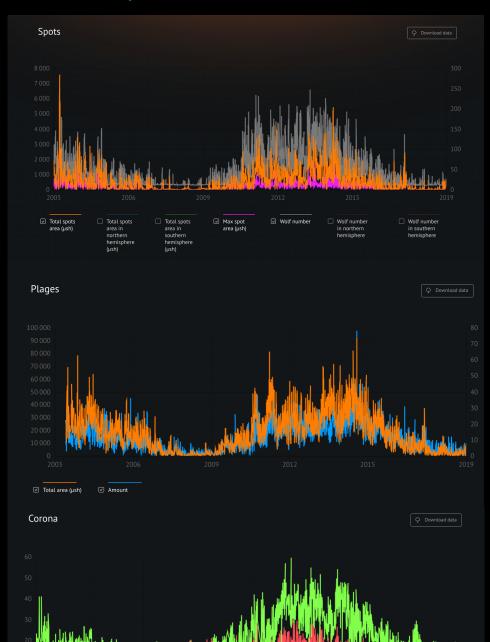
Reference dataset first!

#### Solar activity over 100 years is available in a new website



https://observethesun.com

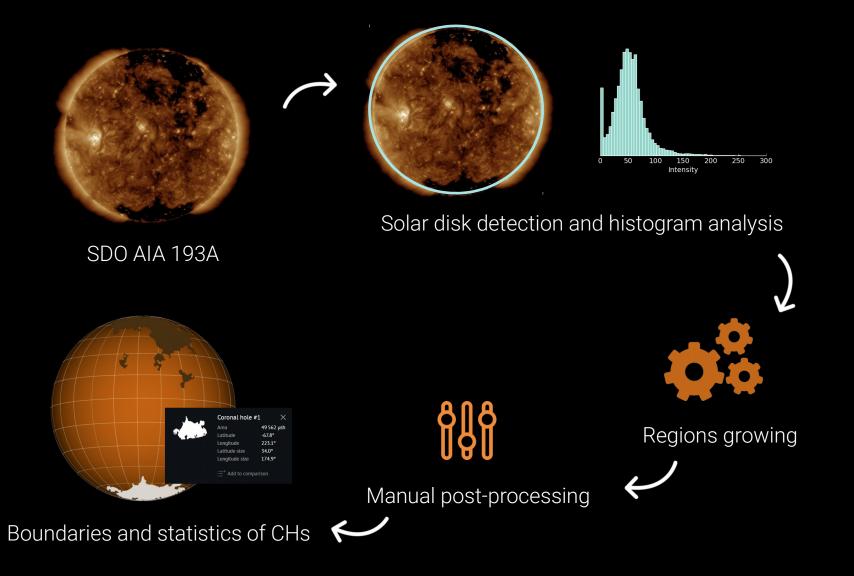
## Solar activity indices at observethesun.com



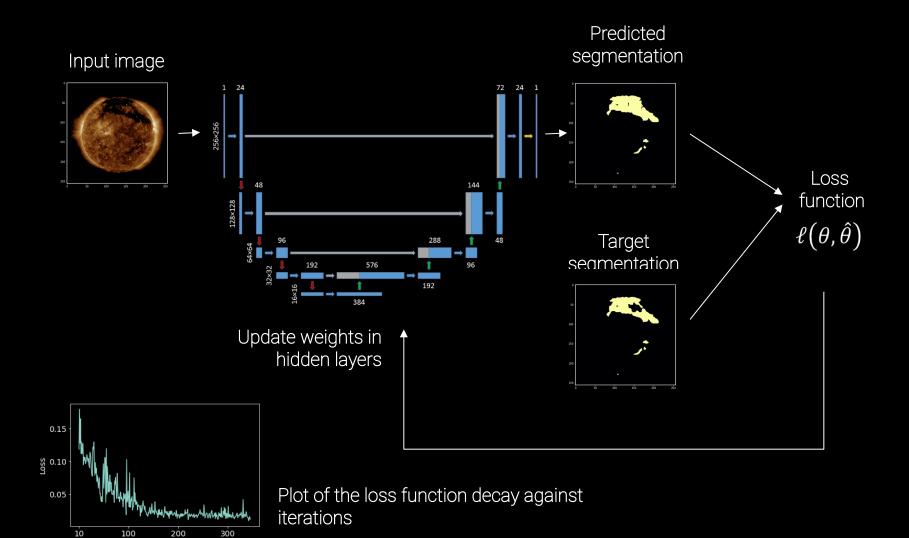
#### Actual forecasts of Kp and solar wind at observethesun.com



Semi-automatic segmentation procedure is applied on a daily basis for production of solar activity maps

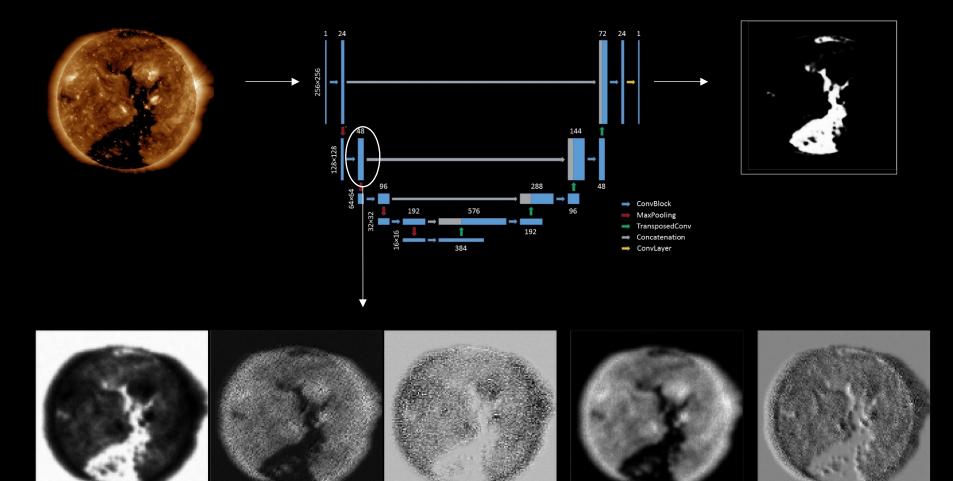


#### Neural network training

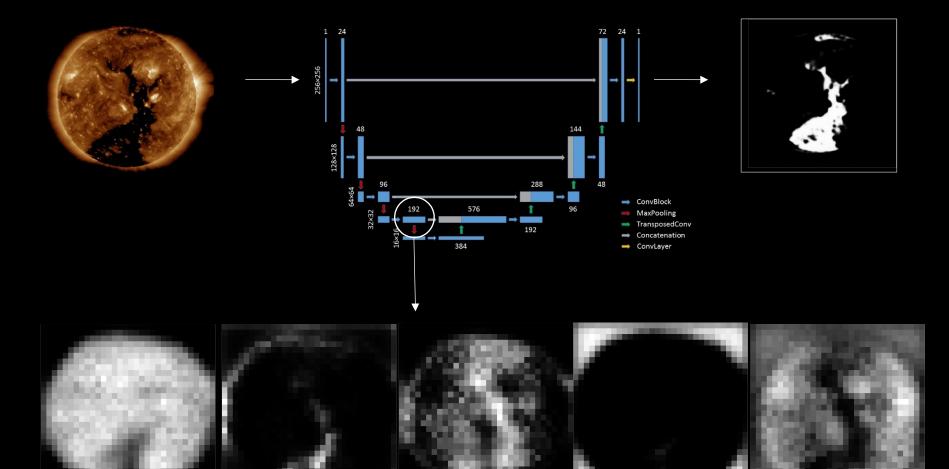


Iterations

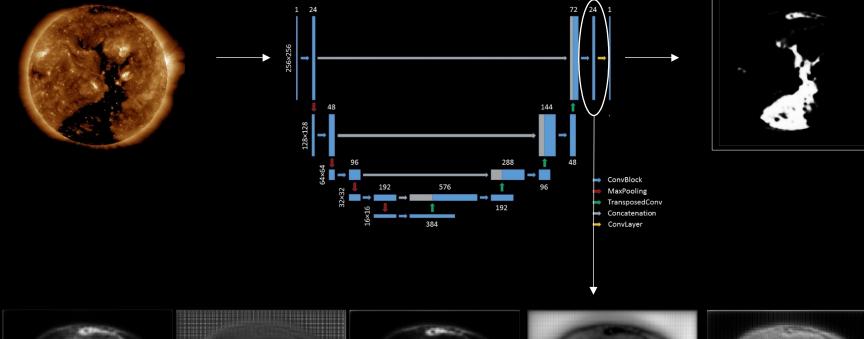
## Hidden layers outputs

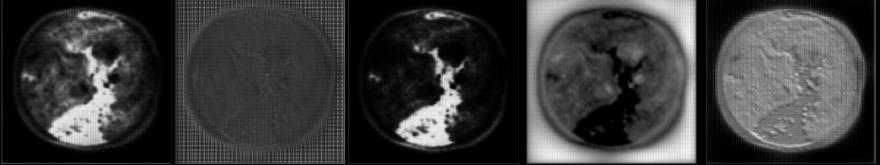


## Hidden layers outputs

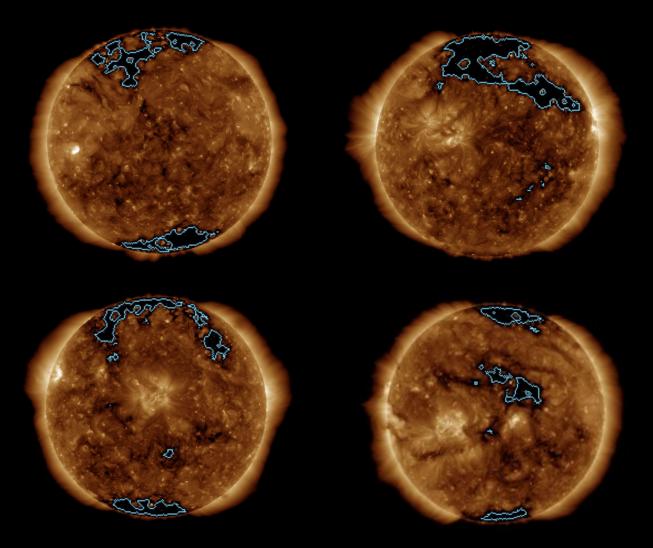


## Hidden layers outputs





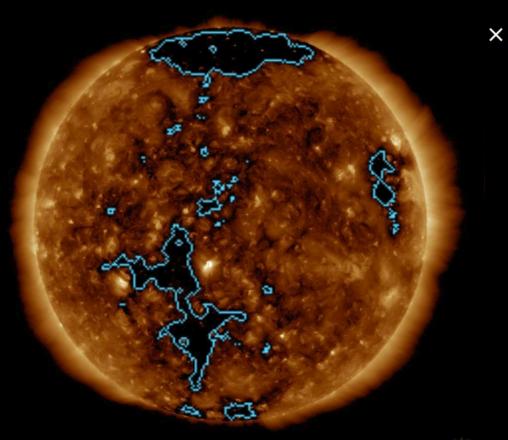
## Coronal holes segmentation with a trained model



Online model for CH segmentation – <u>https://illarionovea.github.io/</u>

# Coronal holes segmentation tool

End-to-end neural network approach



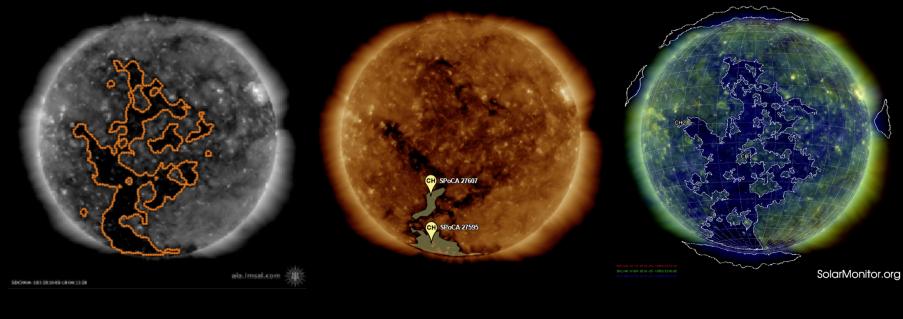


SDO/AIA- 193 2019-08-06 06:58:16

https://illarionovea.github.io/

## Comparison of segmentation maps for the same day

19 March 2018

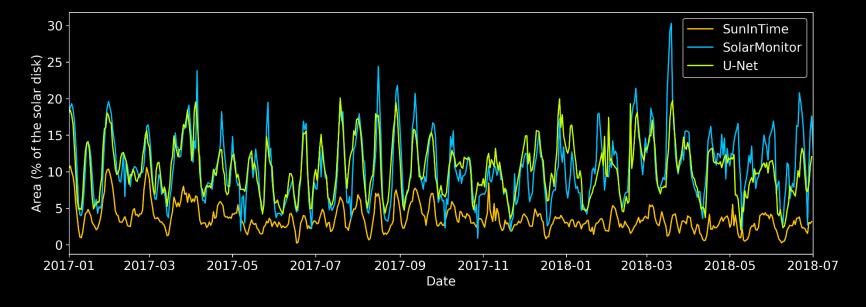


U-Net

Helioviewer

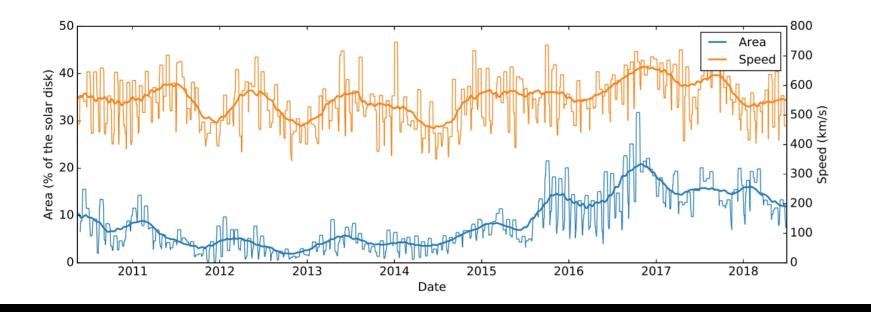
SolarMonitor

#### Total CH's area in comparison to the U-Net model



Correlation with SolarMonitor - 0.76

#### Long-term variations of the total CH's area



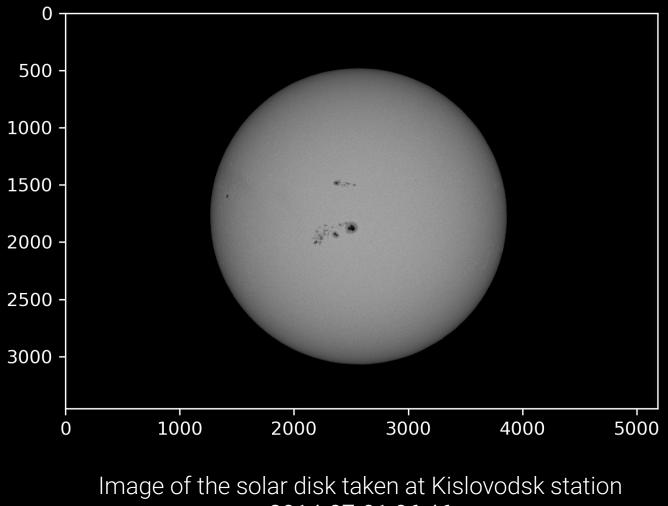
CHs area demonstrates yearly variations, increasing trend during the declining phase of the solar cycle and is minimal during the maximum of the solar cycle. We also note a correlation between CH areas and SW speeds (numerical value of this correlation is 0.7).

E.A. Illarionov, A.G. Tlatov (2018). Segmentation of coronal holes in solar disk images with a convolutional neural network. MNRAS, 481, 4 <u>https://doi.org/10.1093/mnras/sty2628</u>

GitHub repository with source code and practical examples <a href="https://github.com/observethesun/coronal\_holes">https://github.com/observethesun/coronal\_holes</a>

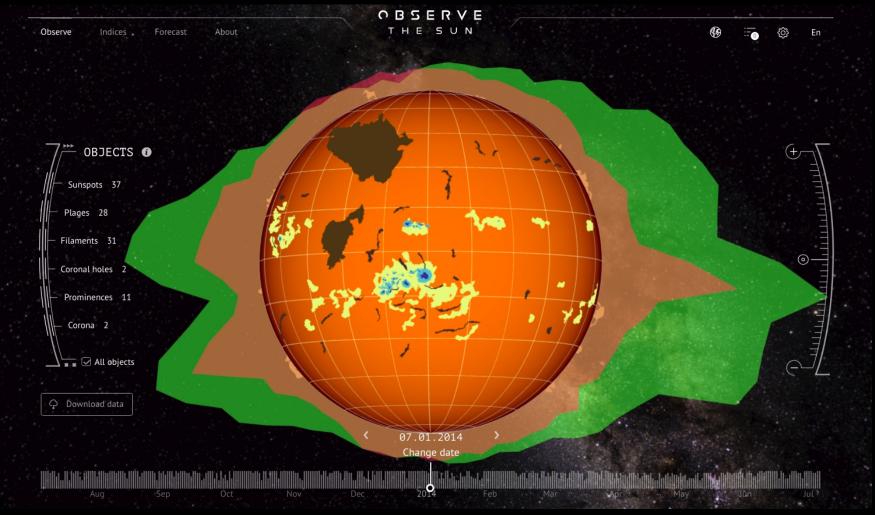
```
def preprocess(batch):
return (batch.load_images(fmt='blosc') #Load source images from blosc file format
    .load_objects() #load coronal holes
    .drop_empty_days() #Drop days without coronal holes
    .make_segmentation_masks((1024, 1024), ohe=False) #Make binary target images
    .downsize_image(['images', 'masks'], 256) #Resize images to 256 x 256
    .random_rot90(['images', 'masks']) #Augmentation
    .random_flip(['images', 'masks'], axis=0) #Augmentation
    .random_flip(['images', 'masks'], axis=1) #Augmentation
    .apply(lambda x: x / 255, ['images']) #Change pixel intensity range to [0, 1]
    .
```

#### Sunspot segmentation

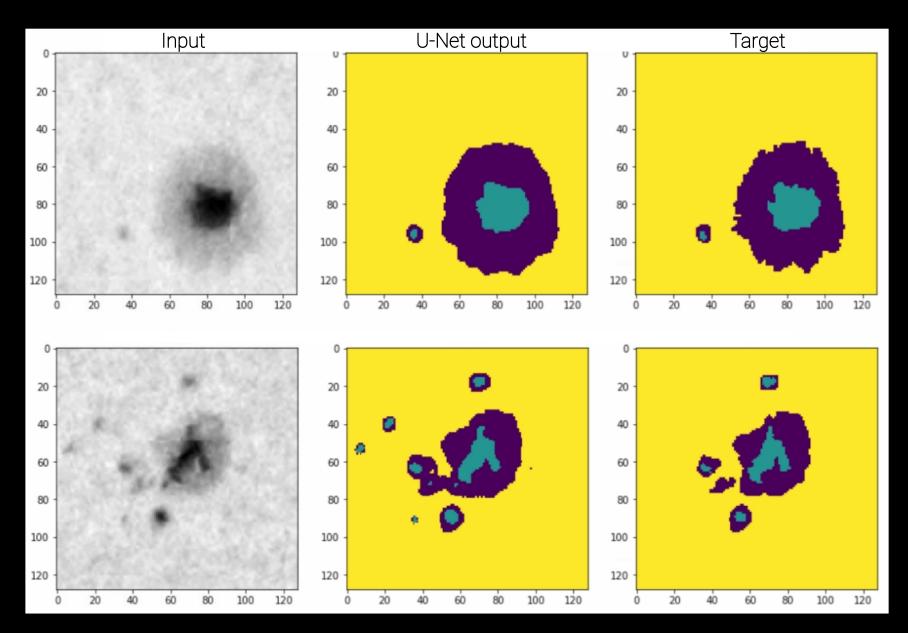


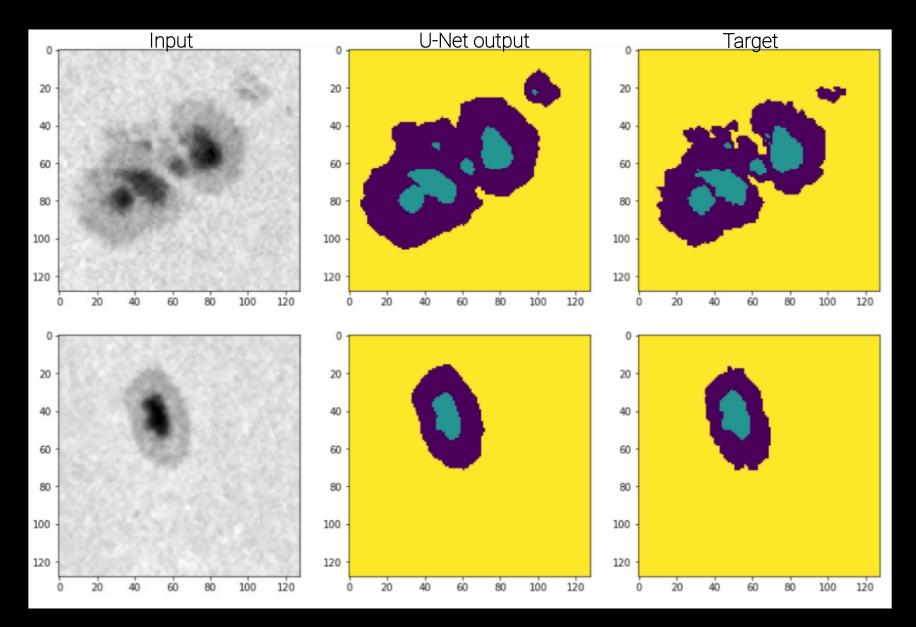
2014-07-01 06:46

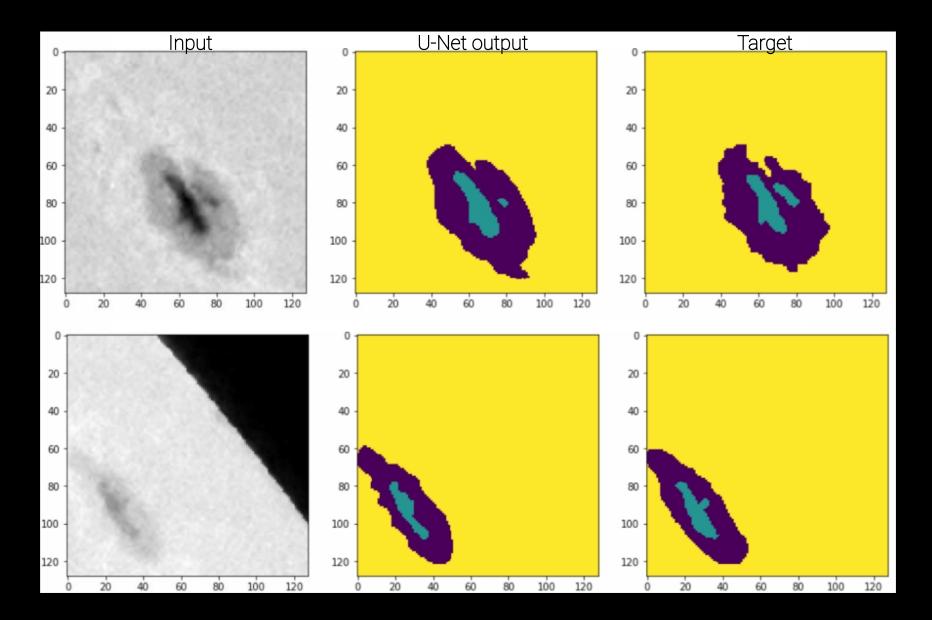
#### Sunspot segmentation

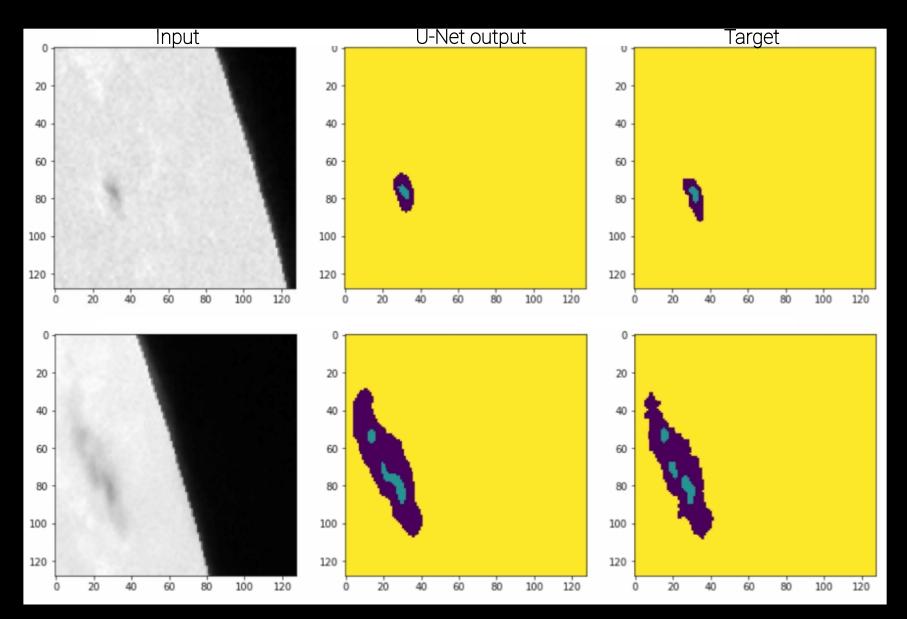


Target segmentation at observethesun.com 2014-07-01

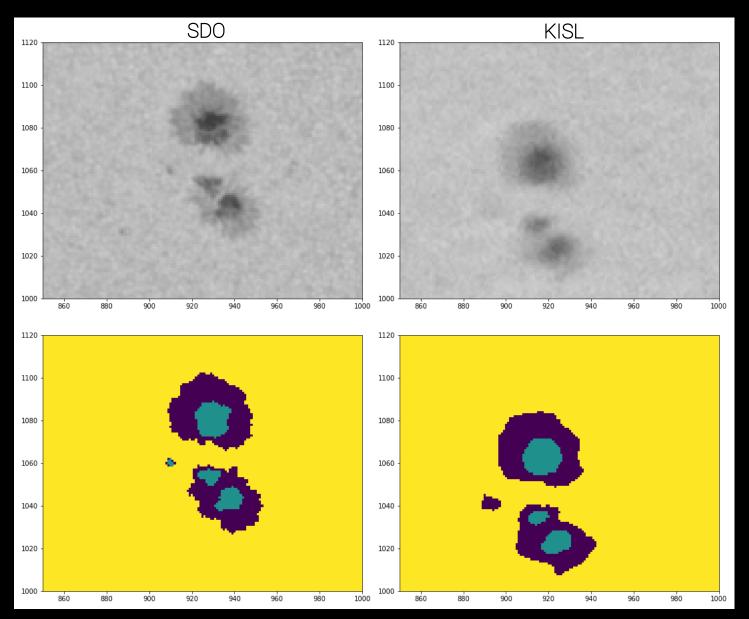




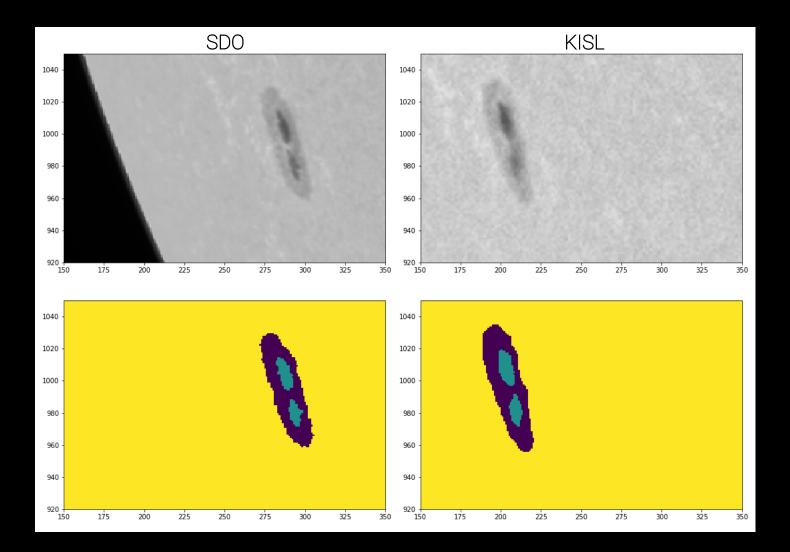




## Application to SDO data



## Application to SDO data



## Conclusions

Conventional automatic image processing tools produce weakly consistent results.

Neural network models are able to produce segmentation similar to manual image interpretation.

Reference datasets are required to compare models and measure progress. Datasets produced by Kislovodsk solar station and available at observethesun.com may be a starting point in this way.

# Thank you for attention!

E. A. Illarionov, A. G. Tlatov (2018). Segmentation of coronal holes in solar disk images with a convolutional neural network. <u>https://doi.org/10.1093/mnras/sty2628</u>

Source code <u>https://github.com/observethesun/coronal\_holes</u>

Online demo <u>https://illarionovea.github.io/</u>

K. Tlatova et al (2018) Reconstruction of Centennial Series of Solar Activity. <u>https://doi.org/10.1134/S0016793218080182</u>

Tlatov,A. G. et al (2016). Forecast of solar wind parameters according to STOP magnetograph observations. <u>https://doi.org/10.1134/S0016793216080223</u>