

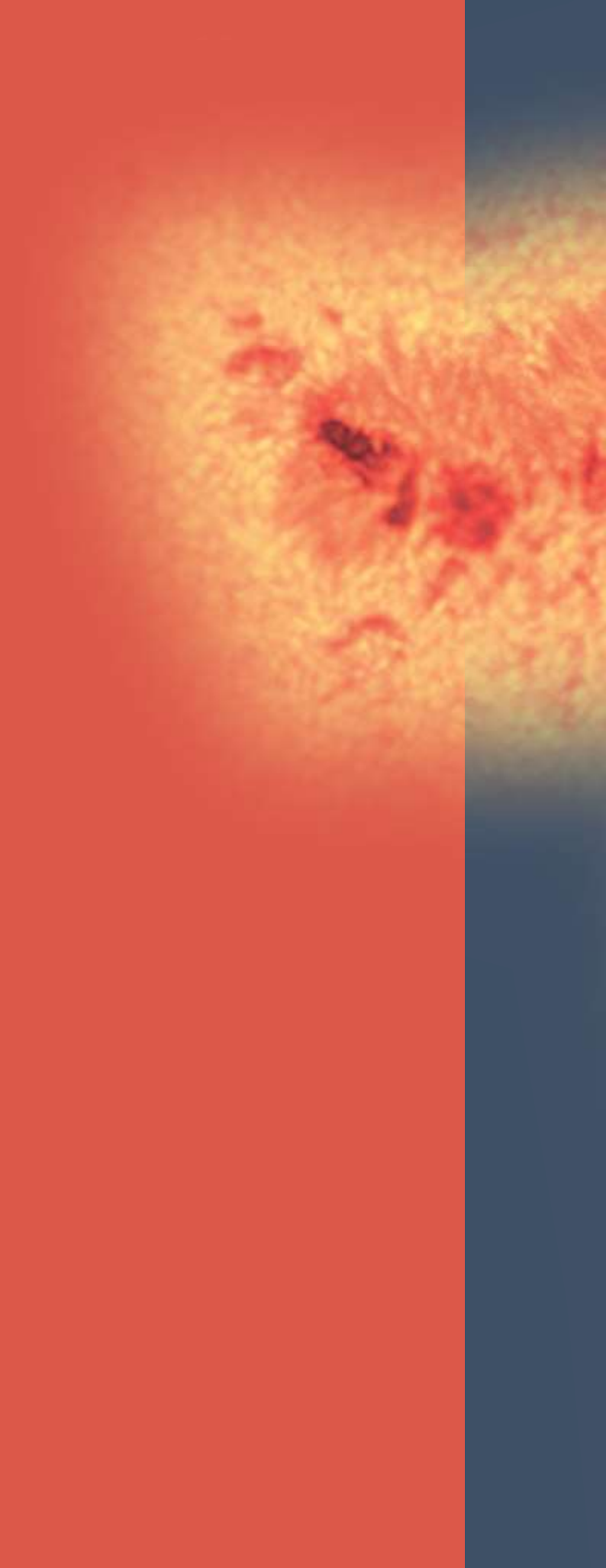
# Sunspot class

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# Abstract

Sunspots, visible through telescopes or with the naked eye, are caused by the Sun's magnetic field. Studying solar activity is crucial for predicting solar activity and understanding the Sun's cycle. My research focuses on automating the recognition, classification, and analysis of sunspot images drawn from the Astronomical Institute of the Czech Academy of Sciences. I trained a convolutional neural network to label sunspots based on the McIntosh classification system, aiming to bridge deep learning with solar astronomy.

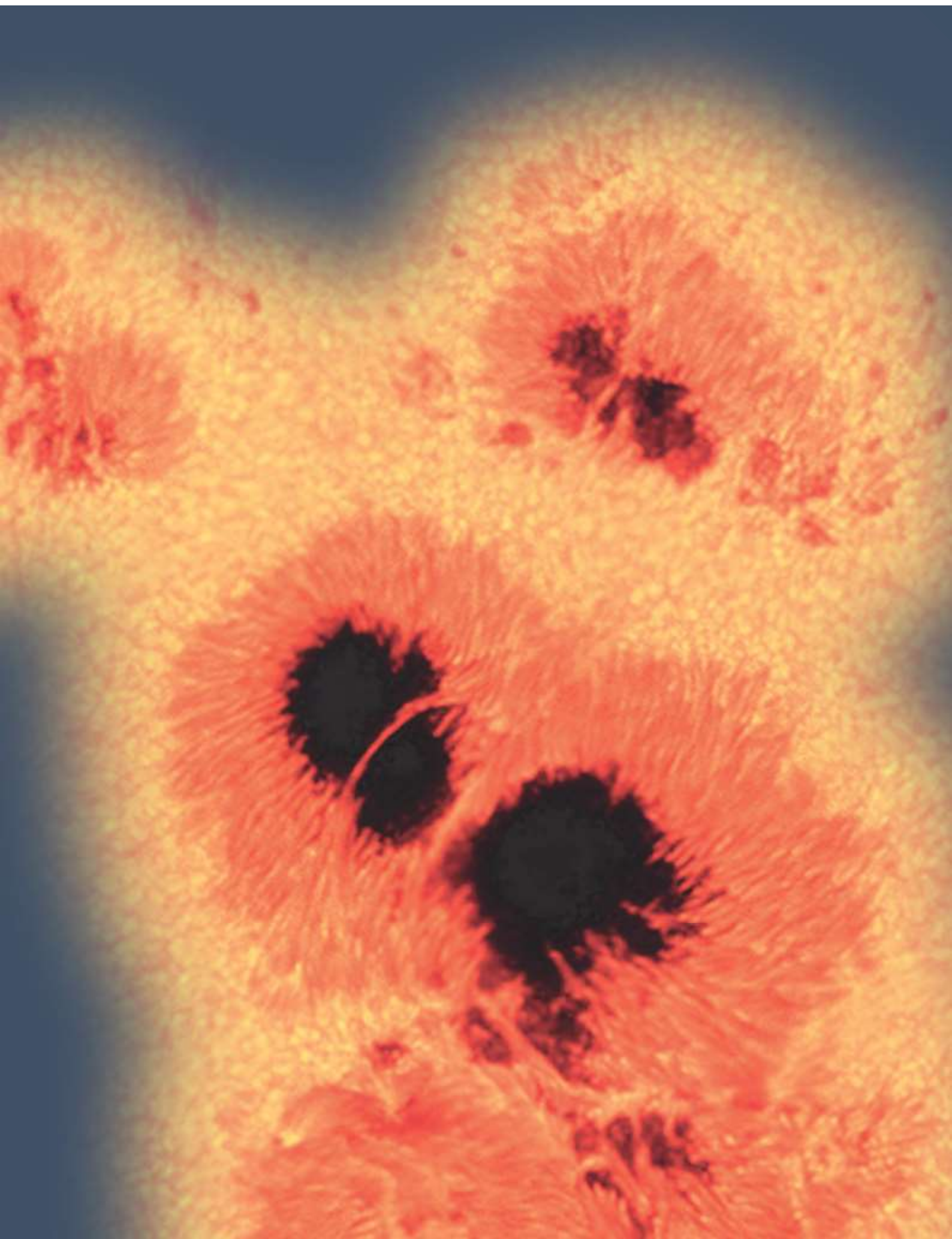
# Methods

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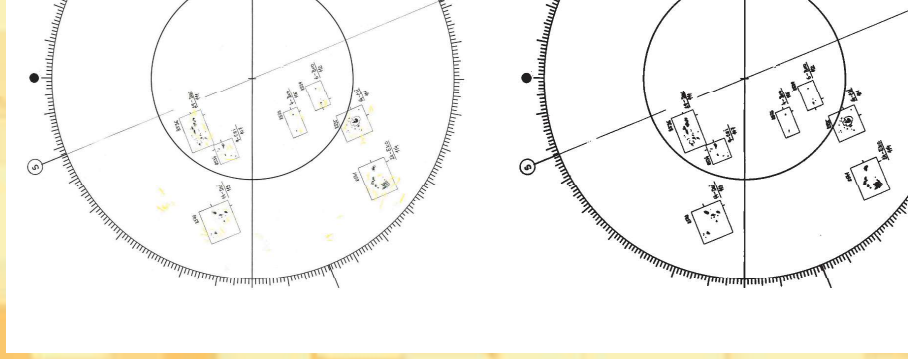
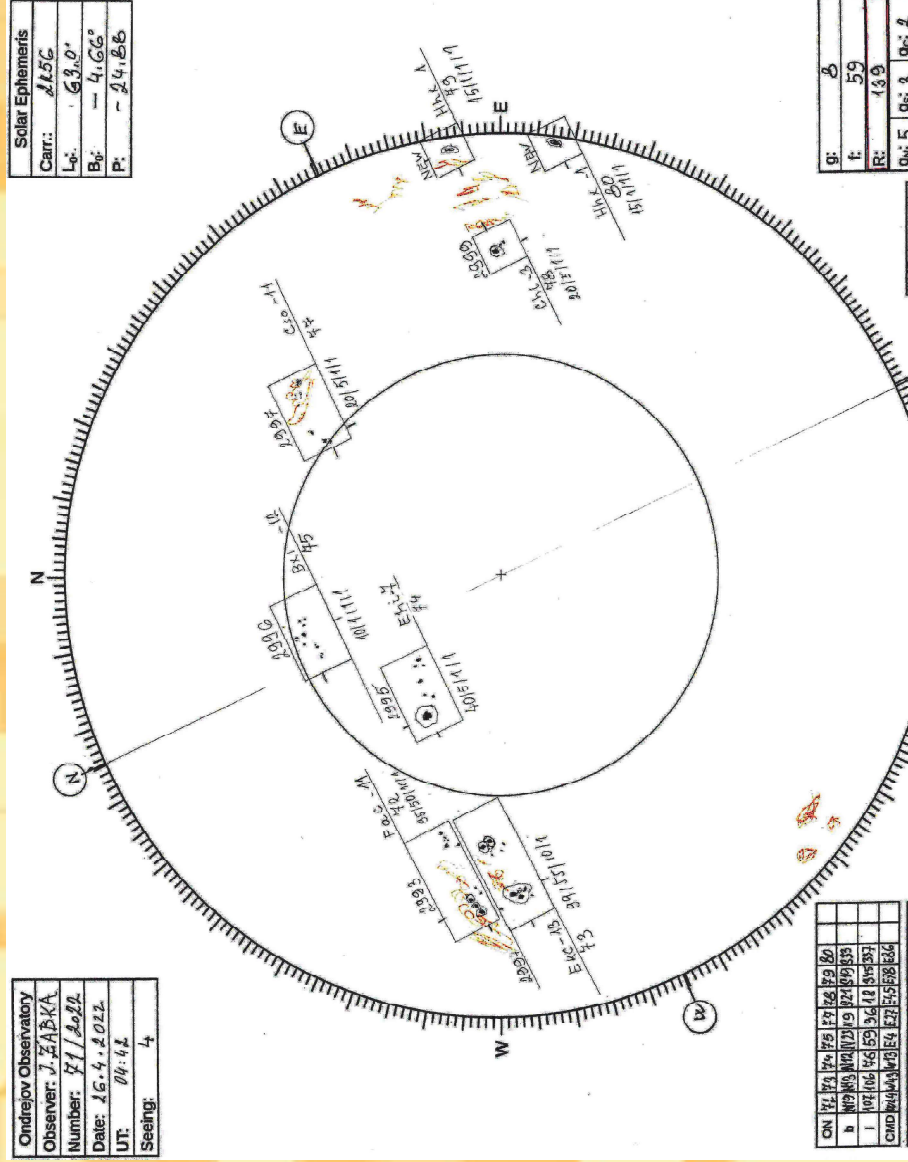
are automated using Python The initial





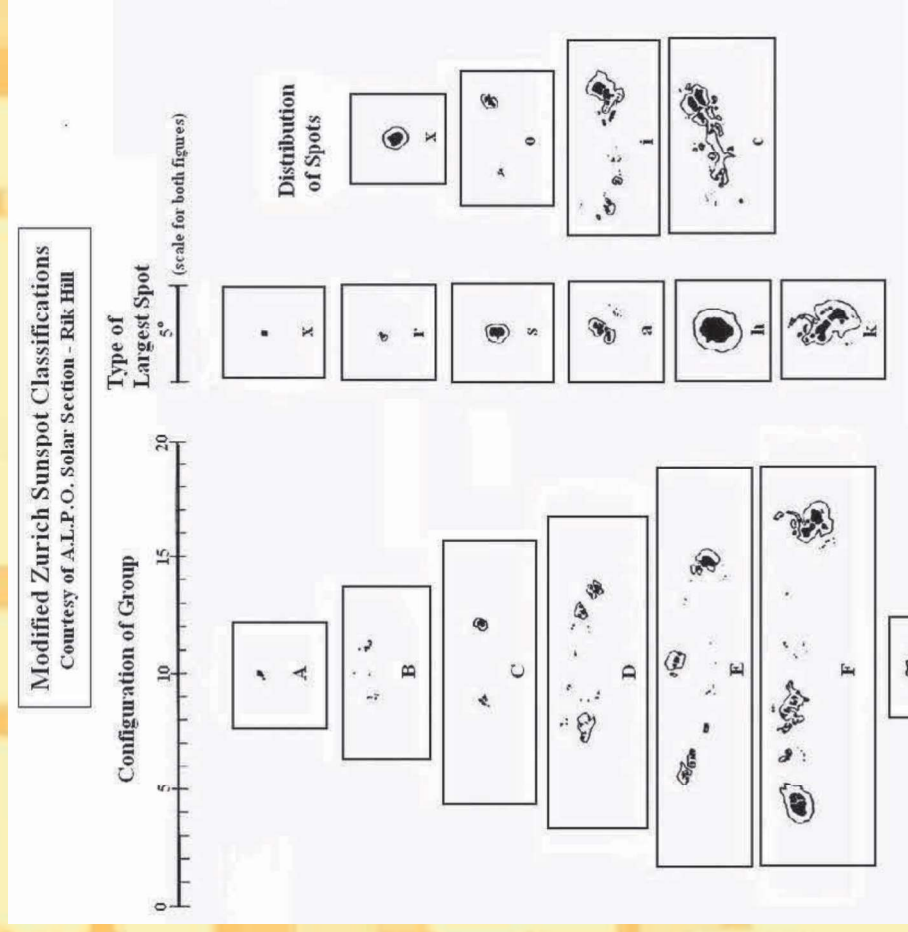
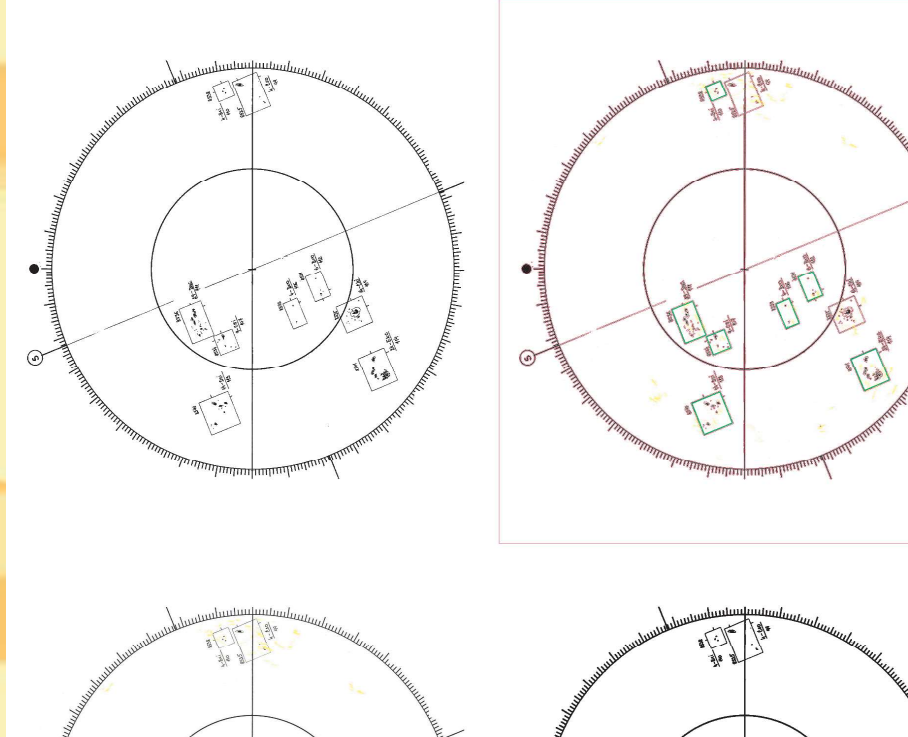
the preprocessing and classification of sunspot data. The dataset consisted of hand-drawn sunspot observations, which were processed through a series of adjustments. The processed images were then used in a classification system, providing key information on sunspot activity.

For model training, the dataset was divided into training and testing sets. A Convolutional Neural Network (CNN) was used to classify the sunspot images. The selection was based on the validation loss function, and the results were evaluated using the test set.



are automated using Python. The initial  
which were prepared for machine learning  
were then labeled according to the McIntosh  
spot size, structure, and distribution.

ing, validation, and test sets in an 80-10-10 ratio. I used a  
spots, utilizing libraries such as Keras and TensorFlow. Model  
using the model that minimized the loss. After training, predictions  
ing accuracy metrics and a confusion matrix.

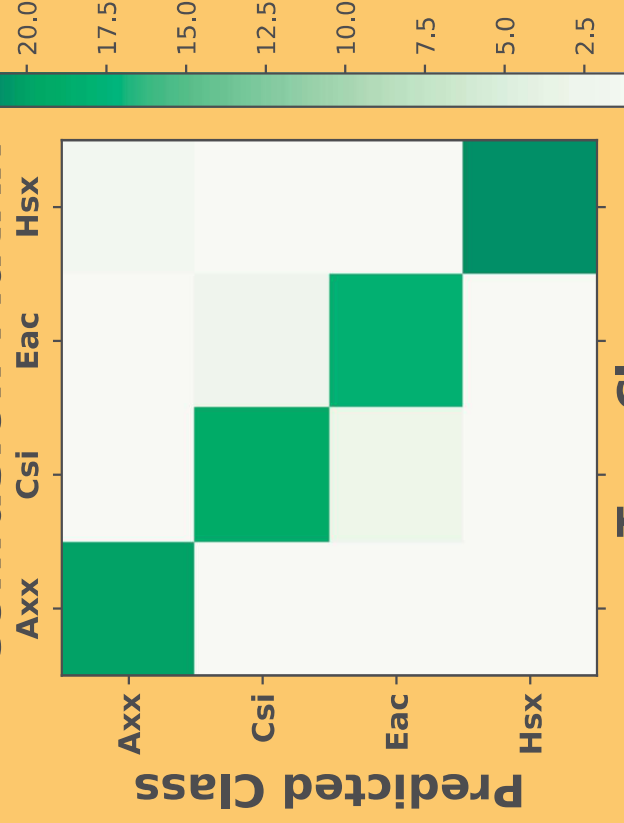




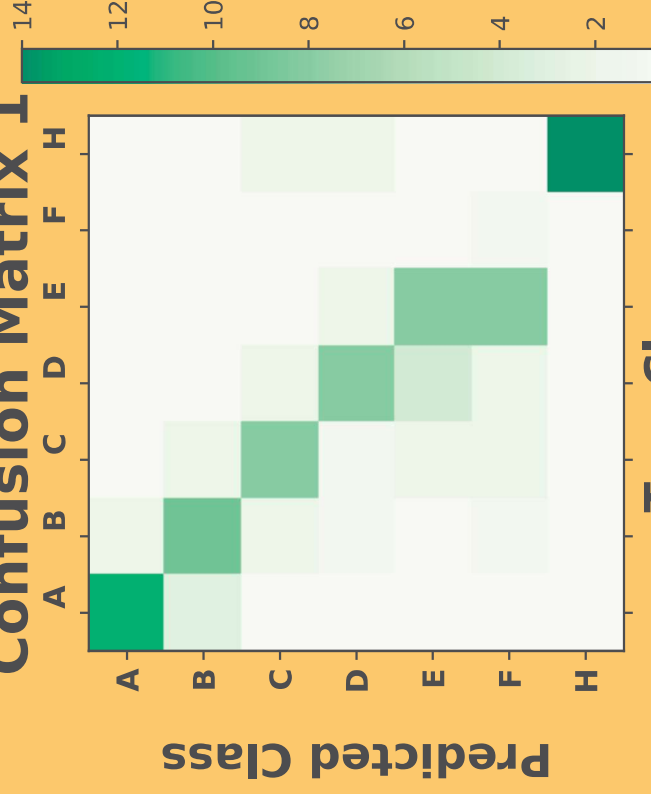
# Results

I've trained several convolutional neural network (CNN) models capable of detecting sunspots at varying levels of complexity, showcasing the potential applications of AI in solar astronomy. Additionally, I have outlined several pathways for future research in this field.

**Confusion Matrix**



**Confusion Matrix 1**



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Whole Disk Forecast			
C	M	X	P
99	55	10	10

F:	L
f <sub>NL04</sub>	f <sub>S</sub> : 35
f <sub>NL04</sub>	f <sub>S</sub> : 5
f <sub>NL04</sub>	f <sub>S</sub> : 19

Whole Disk Forecast

C	M	X	P
99	55	10	10

Whole Disk Forecast

C	M	X	P
99	55	10	10

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Whole Disk Forecast

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99	55	10	10

Whole Disk Forecast

C	M	X	P
99	55	10	10

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CMD

preprocessing: [02a] Key area  
[02b] Removed yellow  
[02c] Enhanced contours  
and black-white conversion  
highlighted detected contours in

Figure 3: Visualisation of each group  
of McIntosh classification system.  
McIntosh system classifies sunspots  
using three letters to indicate size,  
the largest sunspot's properties, and  
group distribution

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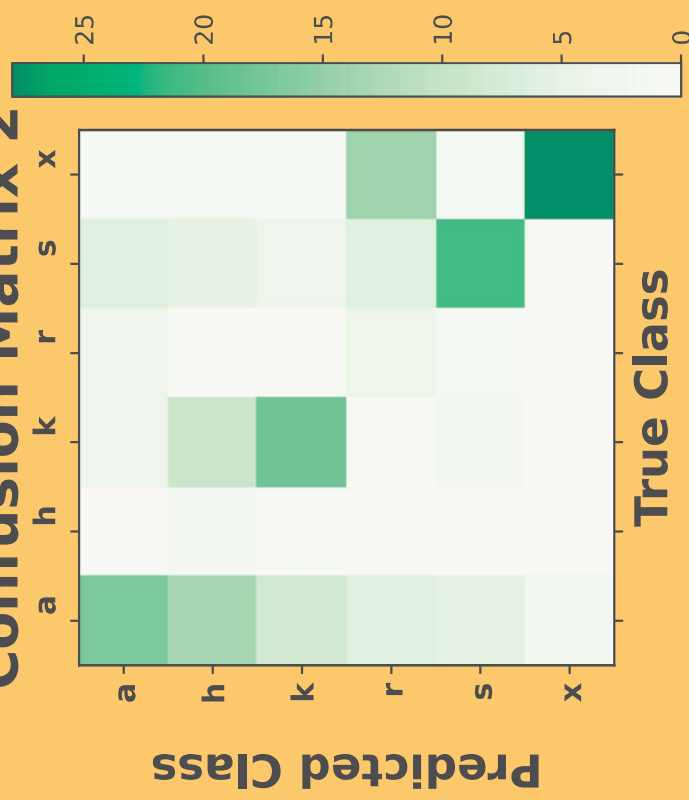


and the square area containing the sunspot. [06b] Removed surrounding areas  
enhance machine learning performance. [06d] Visualisation of labeling for

True Class



Confusion Matrix 2



True Class



Confusion Matrix 3

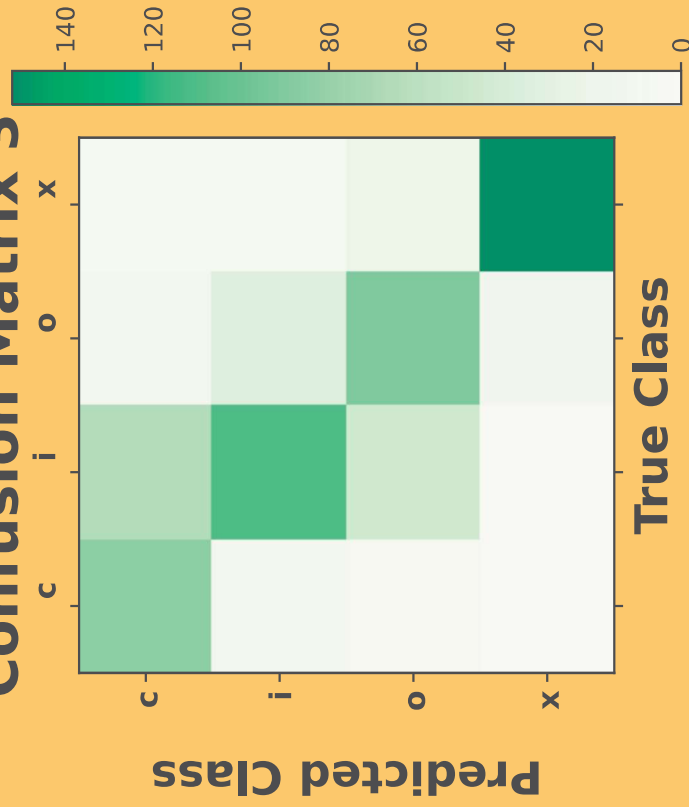


Figure 6: Confusion matrices of selected models. [09a] Model for classification between 4 groups, with 93% accuracy. [09b] Model for predicting the first letter of the McIntosh classification, with 61% accuracy. [09c] Model for predicting the second letter of the McIntosh classification, with 50% accuracy. [09d] Model for predicting the third letter of the McIntosh classification, with 68% accuracy.



# T



Eho



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Figure 5: Visualization of dataset spl

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- [1] KIPPENHAHN, Rudolf. Odhalená tajemství Slunce. Praha
- [2] PROJEKCE - ZÁKRES SLUNEČNÍ FOTOSFÉRY. Astron  
<https://www.asu.cas.cz/~sunwatch/cs/stranka/kresba>

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it into training, validation, and test groups.

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# Future possibilities

Future work could focus on developing more accurate models, improving spot detection, or incorporating data from additional observatories. Expanding model complexity with deeper layers and more neurons could also enhance precision. Another direction is creating a tool for automatic sunspot classification, aiding observatories in verifying manual classifications.



[3] Sunspot SS1 05Sep2016. Online. In: Wikipedia: the 2001-. Dostupné z: <https://upload.wikimedia.org/2024-10-06>].

[4] Astronomický ústav Akademie věd České republiky. Online. In: [www.asu.cas.cz/](http://www.asu.cas.cz/)

[5] nějaký článek/bakalarska s AI?



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ky [online]. c2024 [cit. 2024-10-06]. Dostupné z: <https://>



Figure 7: Sunspot group as drawn at Kanzelhöhe Observatory, Germany.