

MiniSom, a minimalistic and Numpy based implementation of the Self Organizing Maps

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Summary

SOM is a well known type of Artificial Neural Network (Kohonen, 1990) that is able to organize itself so that specific areas respond in a similar way to input patterns thats are similar. Since its first formulation, it has been successfully used for a plethora of applications in many scientific fields. The Machine Learning community has found numerous applications and developed a staggering amount of variants of the original model. MiniSom is a minimalistic and Numpy (Harris et al., 2020) based implementation of the Self Organizing Maps (SOM).

Statement of need

In a scenario where Python has become one of the major languages for scientific development, MiniSom serves three main pruposes. First, offer an implementation of SOM in Python which is easy to use and adapt. Second, give researchers the ability to easily create variants of the main SOM model. Third, offer students an implementation of SOM which is easy to understand.

The interface of MiniSom has evolved to blend with popular Machine Learning frameworks, as scikit-learn (Pedregosa et al., 2011), and the visualization library matplotlib (Hunter, 2007). The documentation of the library is proposed through examples based on ipython notebooks (Pérez & Granger, 2007) and uses the cited libraries.

Applications

At the time I am writing, Minisom has been cited in more than 50 scientific publications. It has been used in many typical Machine Learning applications, as time series modeling (Fortuin et al., 2018) and text mining (Makiyama et al., 2015). And it has also been been used as a tool in a variety of fields, as Geophysics (Lessin et al., 2020) and Climatology (Thompson et al., 2020). MiniSom has been used for the creation of teaching material for courses at University level and MOOCs, see (Ludwig Krippahl, n.d.) for an example of teaching material based on MiniSom.

Historical note

MiniSom was developed while creating a Machine Learning methodology to embed structured data (graphs and trees) into vectorial spaces (Vettigli, 2012; Vettigli & Ciaramella, 2017). The developed has been made while the author was affiliated to institutions 2 and 3.

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Software

- Review 🗗
- Repository 🗗
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