Michael Biehl - Publications

Monographs and Edited Volumes

[1] Michael Biehl. *The Shallow and the Deep: A biased introduction to neural networks and old school machine learning*. University of Groningen Press, 2023. 292 pages.

[2] M. Biehl, B. Hammer, M. Verleysen, and T. Villmann. *Similarity based clustering - recent developments and biomedical applications*, volume 5400 of *Lecture Notes in Artificial Intelligence*. Springer, 2009. 201 pages.

Journal Articles

[3] Alessandro Prete, Katharina Lang, David Pavlov, Yara Rhayem, Alice J Sitch, Anna S Franke, Lorna C Gilligan, Cedric HL Shackleton, Stefanie Hahner, Marcus Quinkler, et al. Urine steroid metabolomics as a diagnostic tool in primary aldosteronism. *The Journal of Steroid Biochemistry and Molecular Biology*, pages Article No. 106445,10 pages, 2023.

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