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## **Efficient morphological tools for astronomical image processing**

Moschini, Ugo

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# **Efficient Morphological Tools for Astronomical Image Processing**

**Ugo Moschini**



Netherlands Organisation for Scientific Research

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Cover background: VISTA gigapixel mosaic of the central parts of the Milky Way. Id: eso1242. Credit: ESO/VVV Survey/D. Minniti. Acknowledgement: Ignacio Toledo, Martin Kornmesser.

Front cover detail: photo of a tea plantation taken by the author in Darjeeling, India. The whole plantation is made of small tea shrubs: likewise, the parallel construction of a max-tree occurs by merging together the small sub-trees computed by the threads.

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# Efficient Morphological Tools for Astronomical Image Processing

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by

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to my grandma Nonna Vera,  
and her beloved Firenze.



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