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## Computer-aided detection of wall motion abnormalities in cardiac MRI

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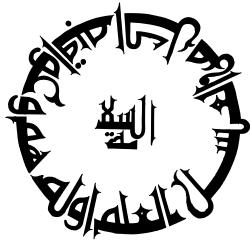
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# Computer-Aided Detection of Wall Motion Abnormalities in Cardiac MRI

Avan Suinesiaputra

## Colophon

About the cover:



The cover is an Arabic calligraphy made by an 11-year old M. Zulkhairi Rhiza Tala, which was based on a Kufic inscription written on an 11th–12th century Iranian dish. The text on the dish is translated into: “Science is bitter-tasting (of thinking) at first, but sweeter than honey in the end. Peace.” The dish is currently stored in the Islamic Art collection of the Musée du Louvre, Paris.

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# Computer-Aided Detection of Wall Motion Abnormalities in Cardiac MRI

## **Proefschrift**

ter verkrijging van  
de graad van Doctor aan de Universiteit Leiden,  
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volgens besluit van het College voor Promoties  
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Avan Suinesiaputra  
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in 1974

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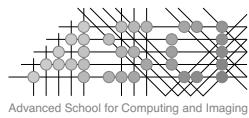
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*Papah/Bapak and Mamah/Mamak,  
from whom I receive most of Allah's love through.*



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