

AutoAI (WP 7)

Holger H. Hoos
(theme leader)

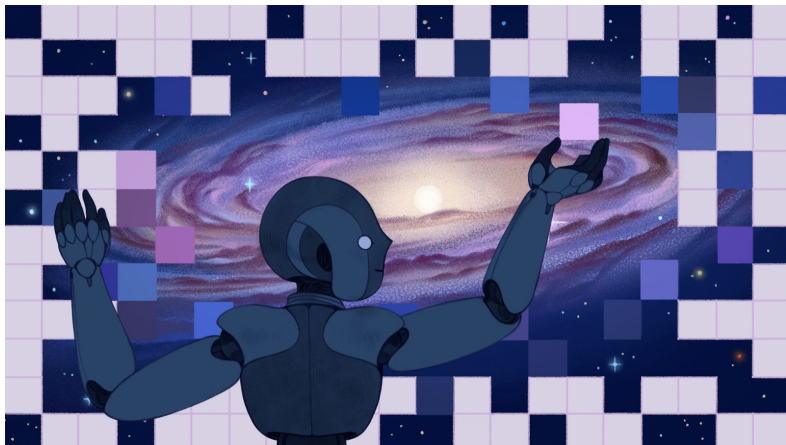
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Universiteit Leiden
The Netherlands

CS Department
University of British Columbia
Canada

TAILOR Workshop
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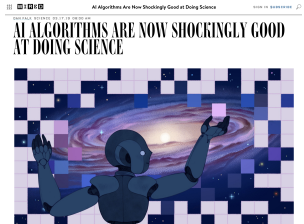
DAN FALK SCIENCE 03.17.19 08:00 AM

AI ALGORITHMS ARE NOW SHOCKINGLY GOOD AT DOING SCIENCE



Shockingly good at doing science ...

“Whether probing the evolution of galaxies
or discovering new chemical compounds,
algorithms are detecting patterns
no humans could have spotted.”



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\rightsquigarrow increased benefits, reduced risks from use of AI

Examples:

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- ▶ Ever-learning AutoAI (WP 7.5; lead: TU Eindhoven)

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 - ▶ AI-on-demand platform

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