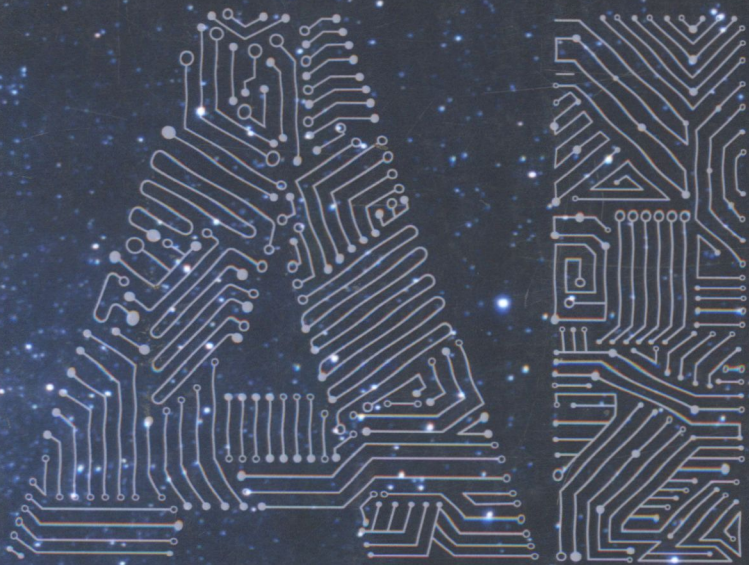


VOLKER KNECHT

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Mario Campanelli • Chao Fang • Jimena González Lozano
Mayank Agrawal • Daniel Grün • Yang-Hui He



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AI FOR PHYSICS



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"Artificial intelligence is on a rapid path of revolutionizing science, upsetting long-established routines, and learning complex relationships in nature beyond what human minds can comprehend. *AI for Physics* hits the nail on the head in explaining how that is possible and where the future may take us. A great read for scientists and anybody else curious about how soon computers will be smarter than us."

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