Real-time EFT with application to QCD matter and perhaps Al





The Nobel Prize in Physics 2024

The Royal Swedish Academy of Sciences has decided to award the Nobel Prize in Physics 2024 to

John J. Hopfield

Geoffrey Hinton

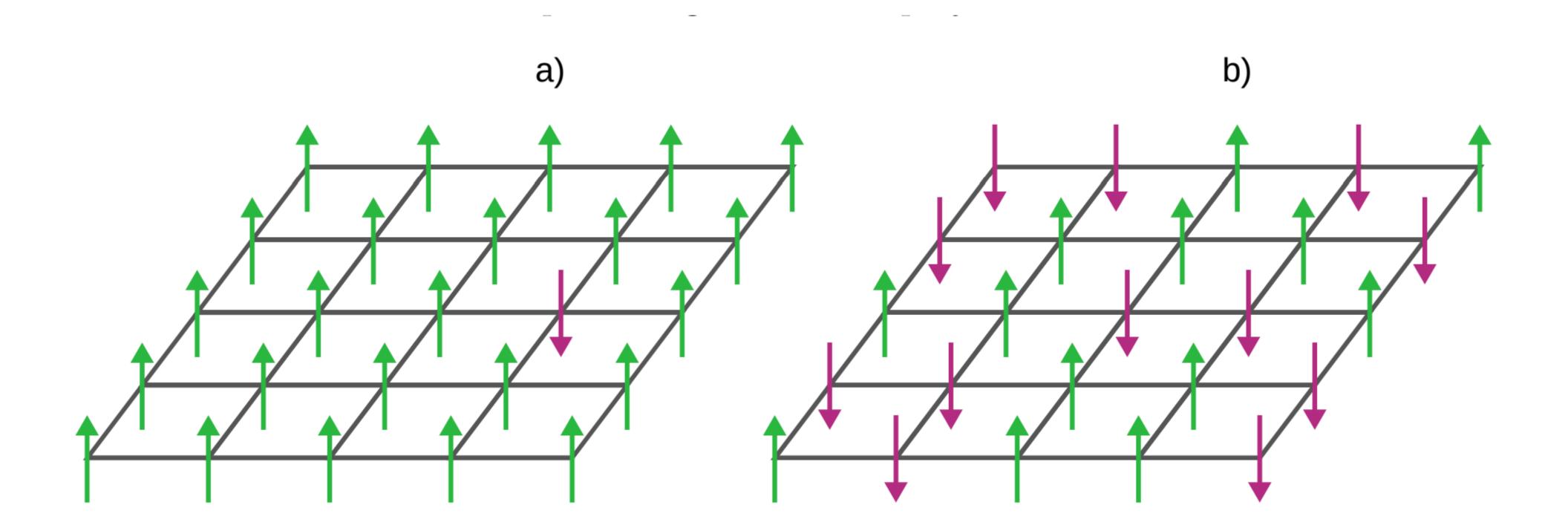
Princeton University, NJ, USA

University of Toronto, Canada

"for foundational discoveries and inventions that enable machine learning with artificial neural networks"

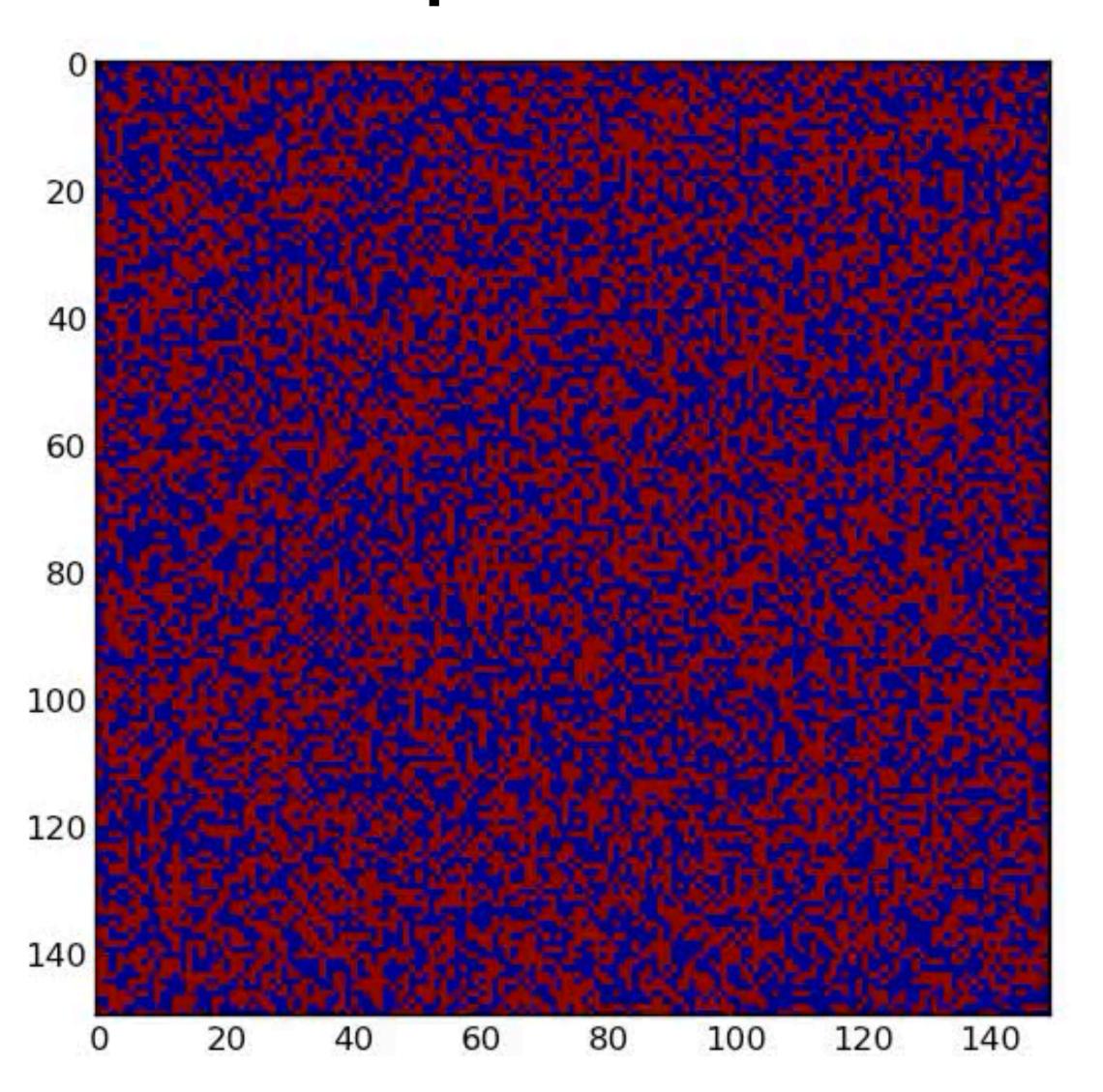
They trained artificial neural networks using physics

2d Ising Model



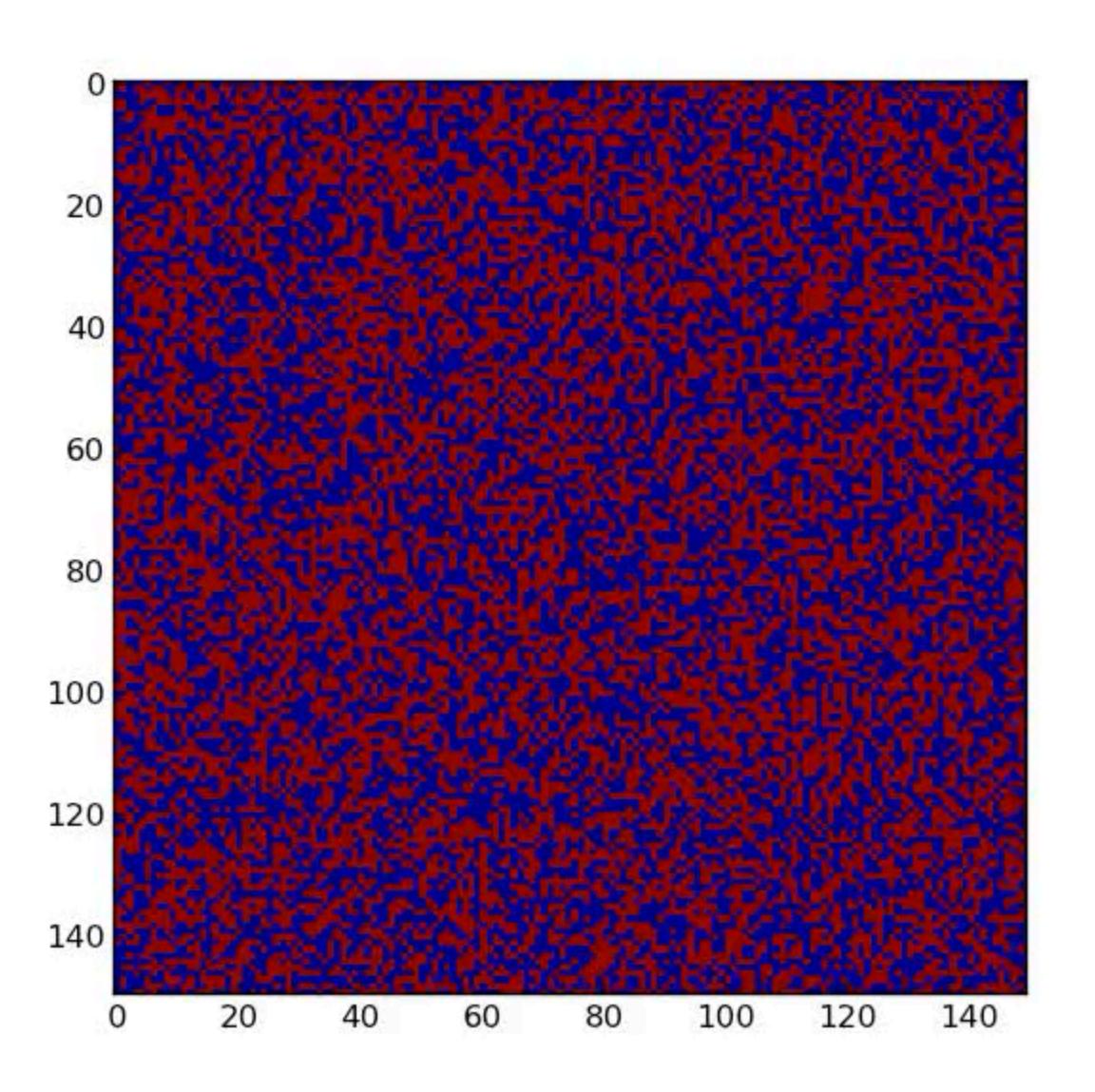
by B.D.Hammel

Low Temperature Phase

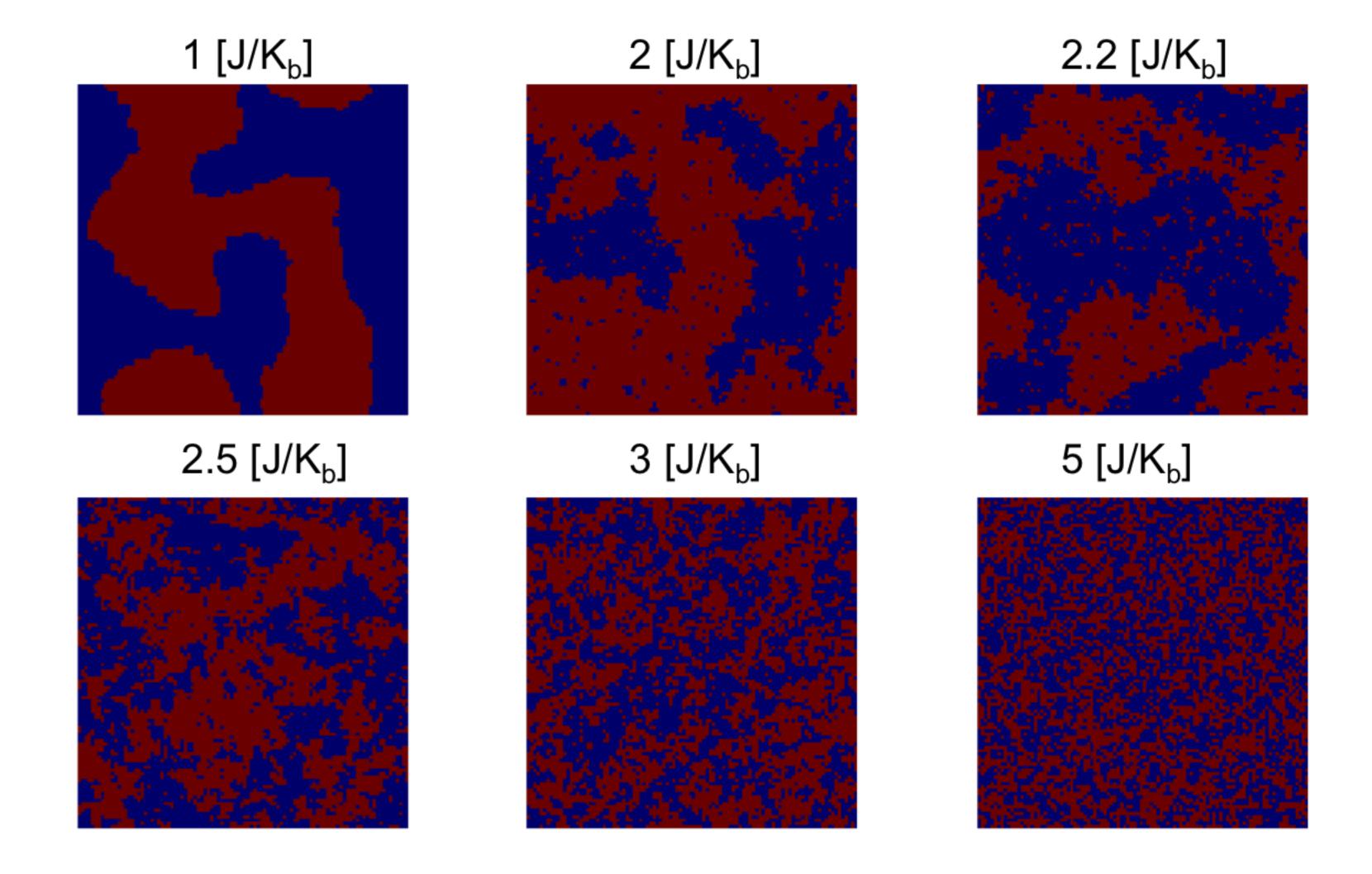


by B.D.Hammel

High-Temperature Phase

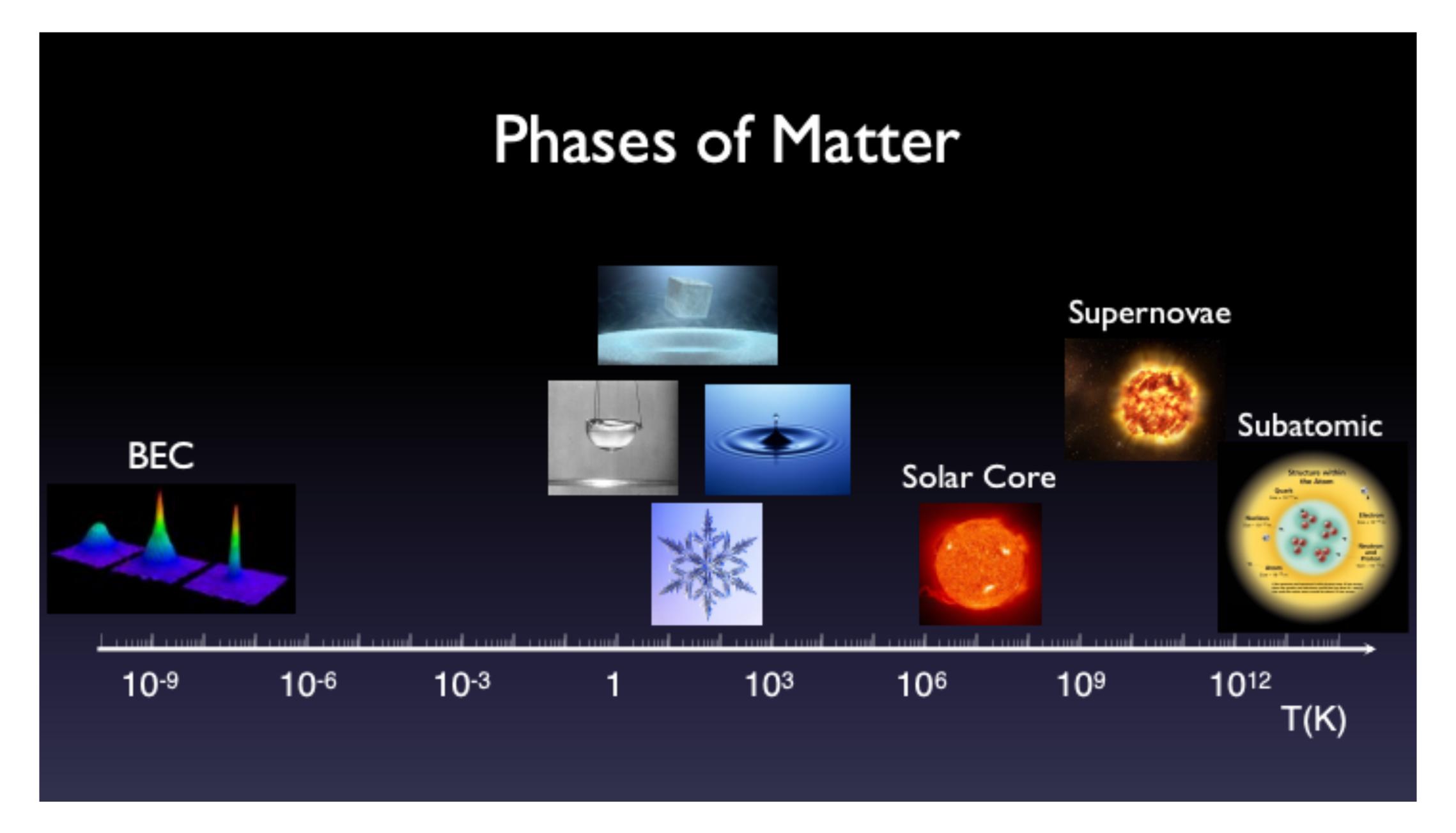


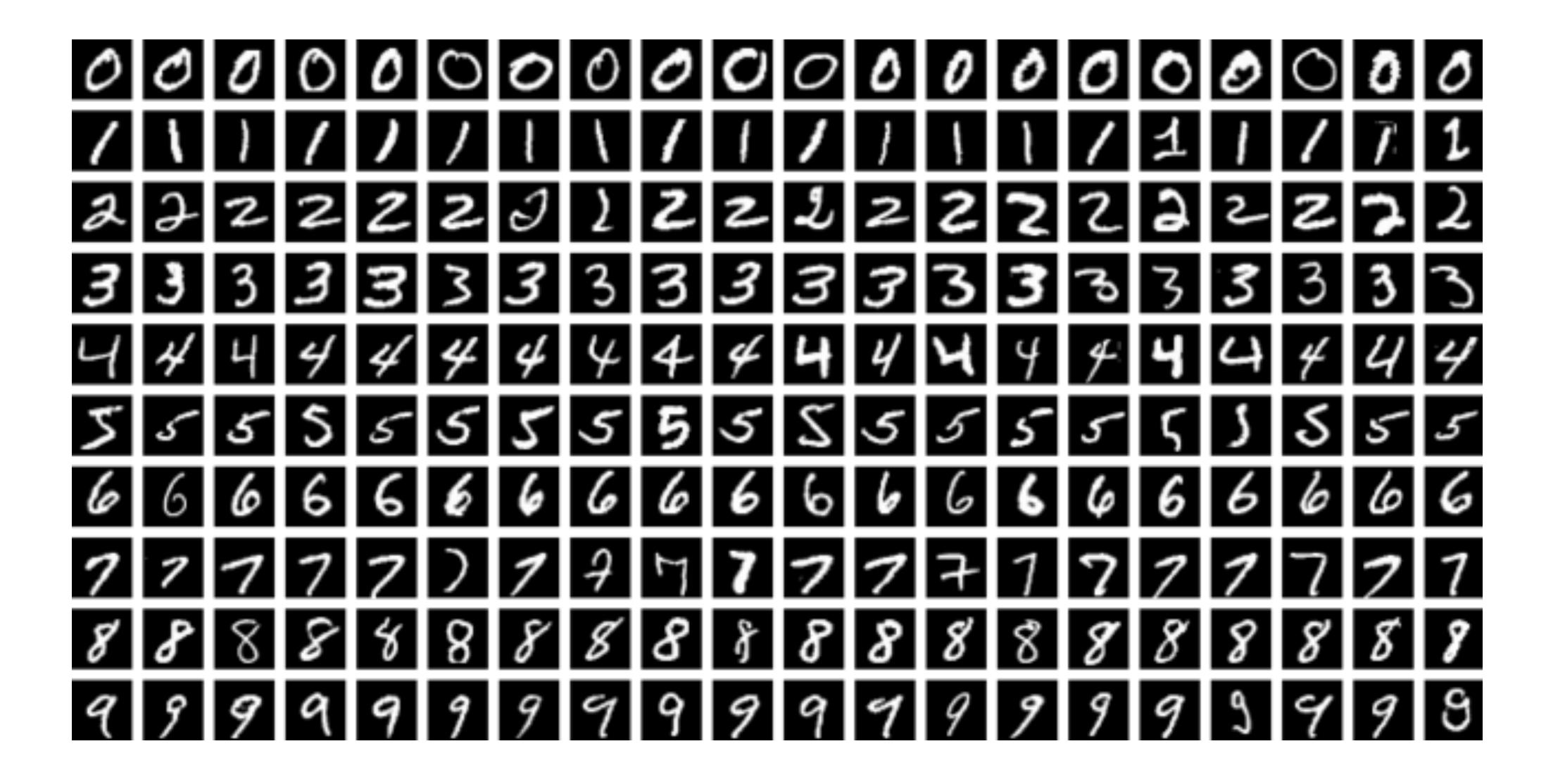
by B.D.Hammel



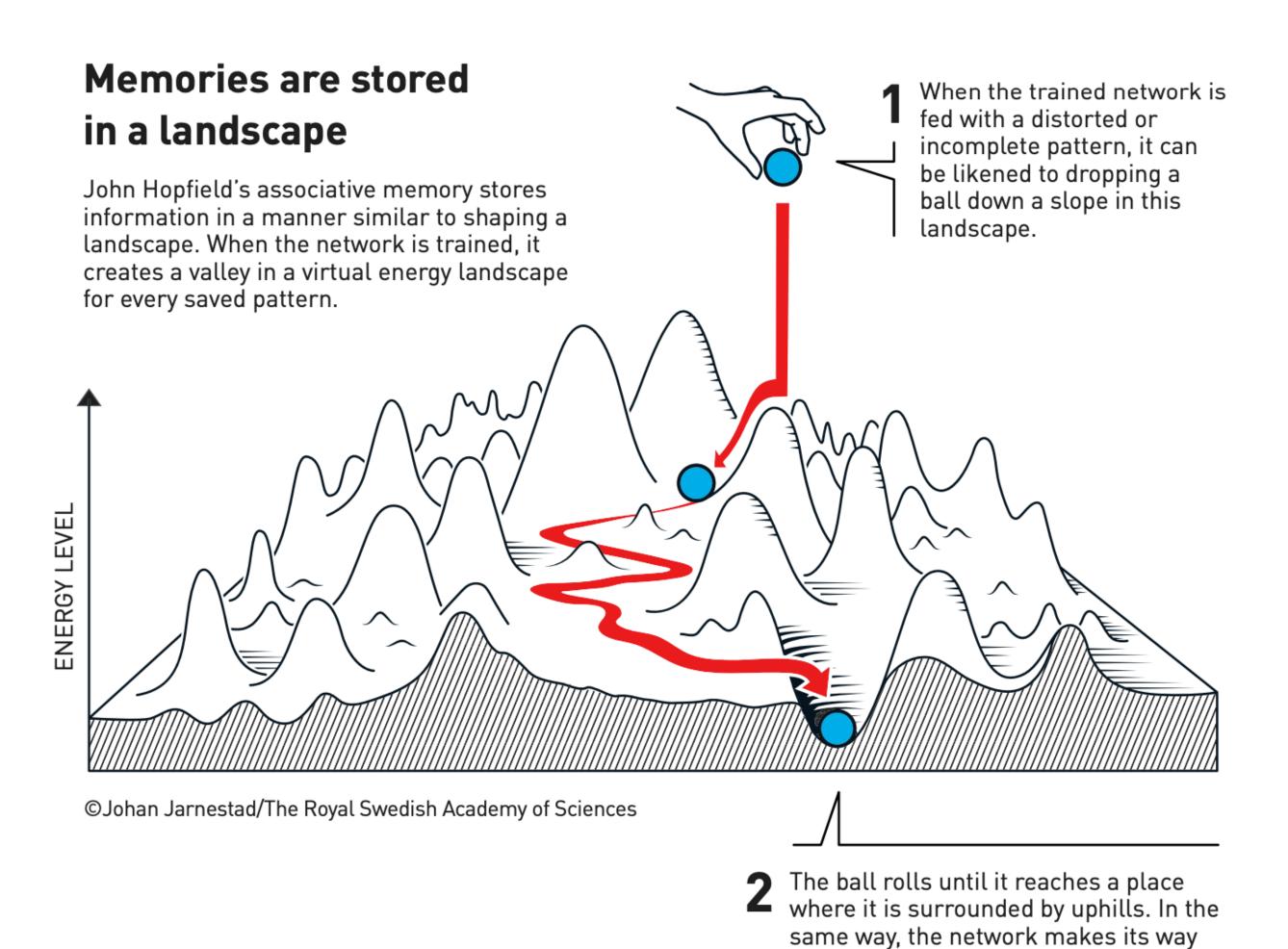
Statistical Physics Inspired Al

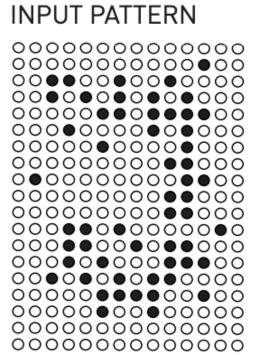
• Leveraging the phase dynamics of artificial particles (neurons, nodes, or units) to model and manipulate patterns for learning, recognition, and generative processes



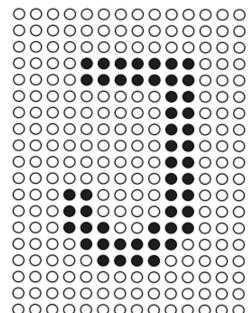


Sample images from MNIST test dataset





SAVED PATTERN

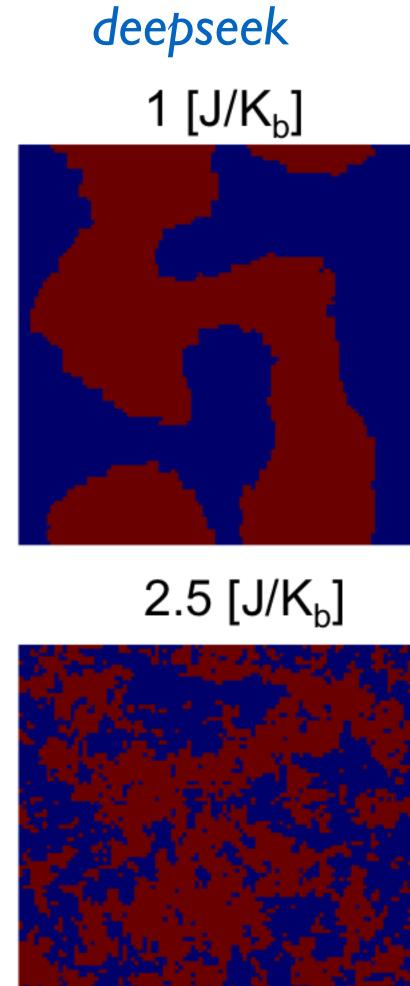


towards lower energy and finds the

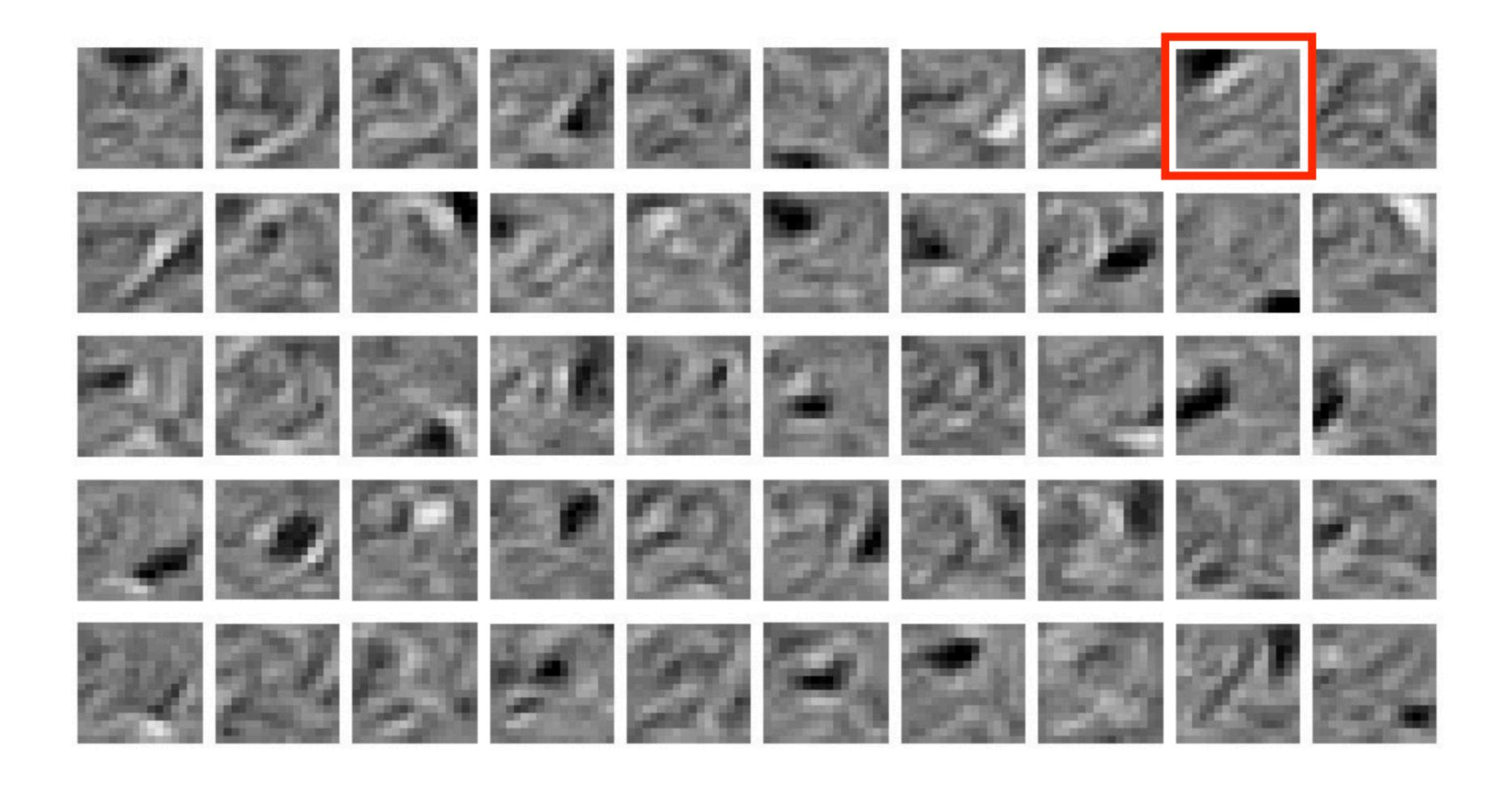
closest saved pattern.

What is your favorite drink?

温度值	回答风格	示例回答
低温 (0.1- 0.3)	严谨、客观	"作为AI,我没有个人偏好,但可以推荐水、咖啡或茶。"
中温 (0.5- 0.7)	自然、略带互动	"虽然我没有味觉,但很多人喜欢冰爽的柠檬茶。你呢?"
高温 (0.8- 1.0)	创意、幽默	"我幻想喝银河系星光奶茶!现实点的话…可能是魔法咖啡?"
超高温度 (1.5)	疯狂、无逻辑	"最爱火山岩浆特饮配彗星尾粉末!不过建议你先试试氧气 泡泡水~"

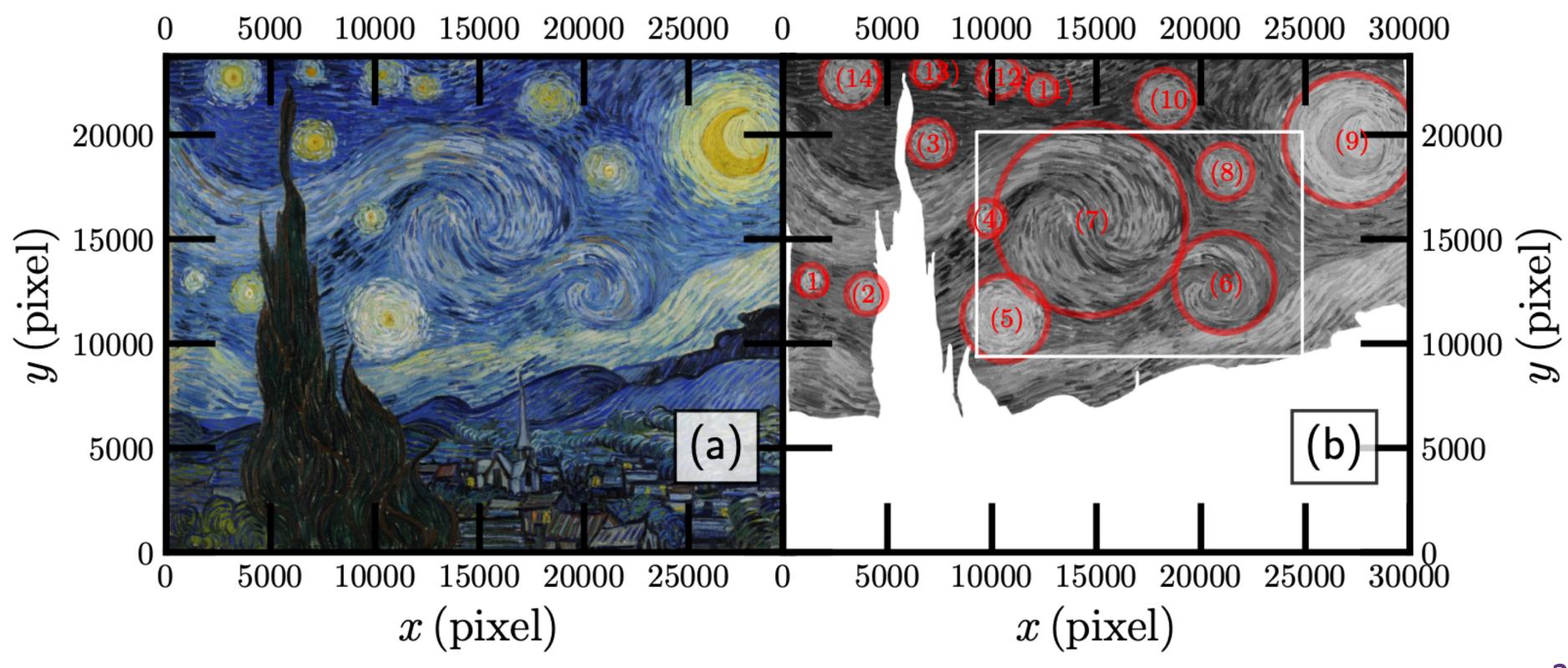


Hidden unit as a feature detector



Part II: Non-equilibrium Phases

The Starry Night and Turbulence



Diffusive model for Al Image Generation

Deep Unsupervised Learning using Nonequilibrium Thermodynamics

Jascha Sohl-Dickstein

Stanford University

Eric A. Weiss

University of California, Berkeley

Niru Maheswaranathan

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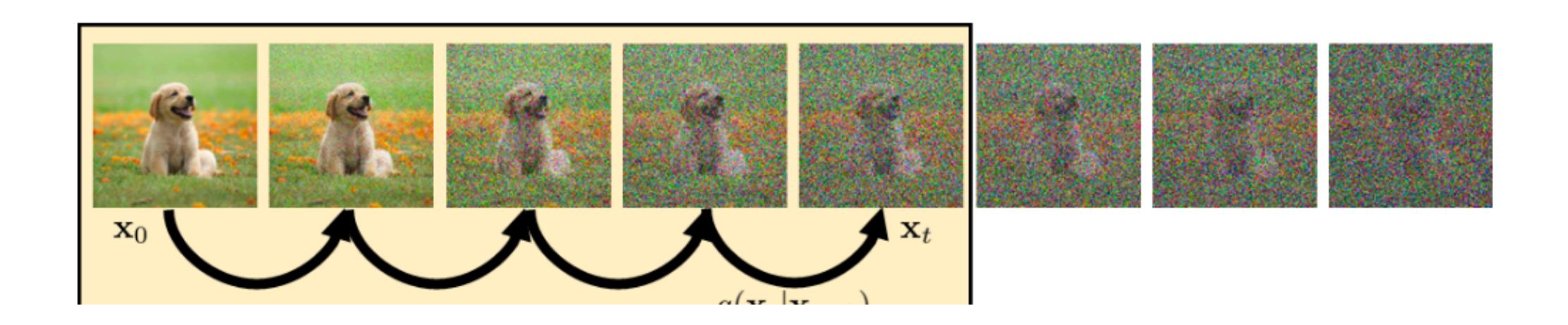
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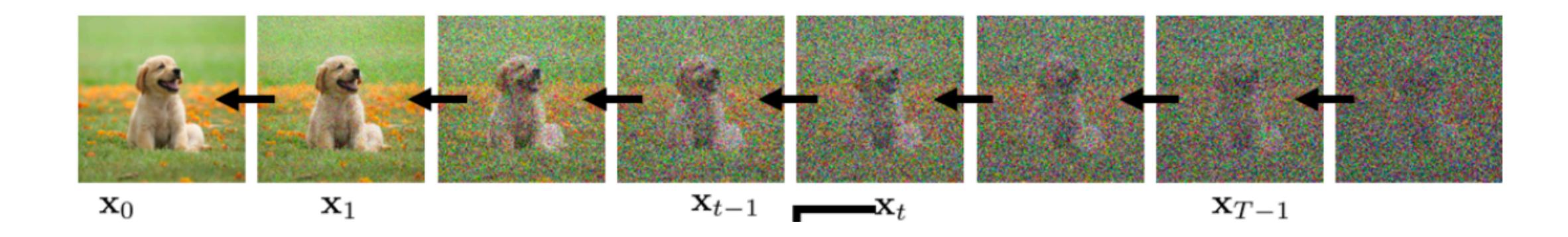
SGANGULI@STANFORD.EDU

• Diffusion models are widely used for image denoising, image generation, and video generation

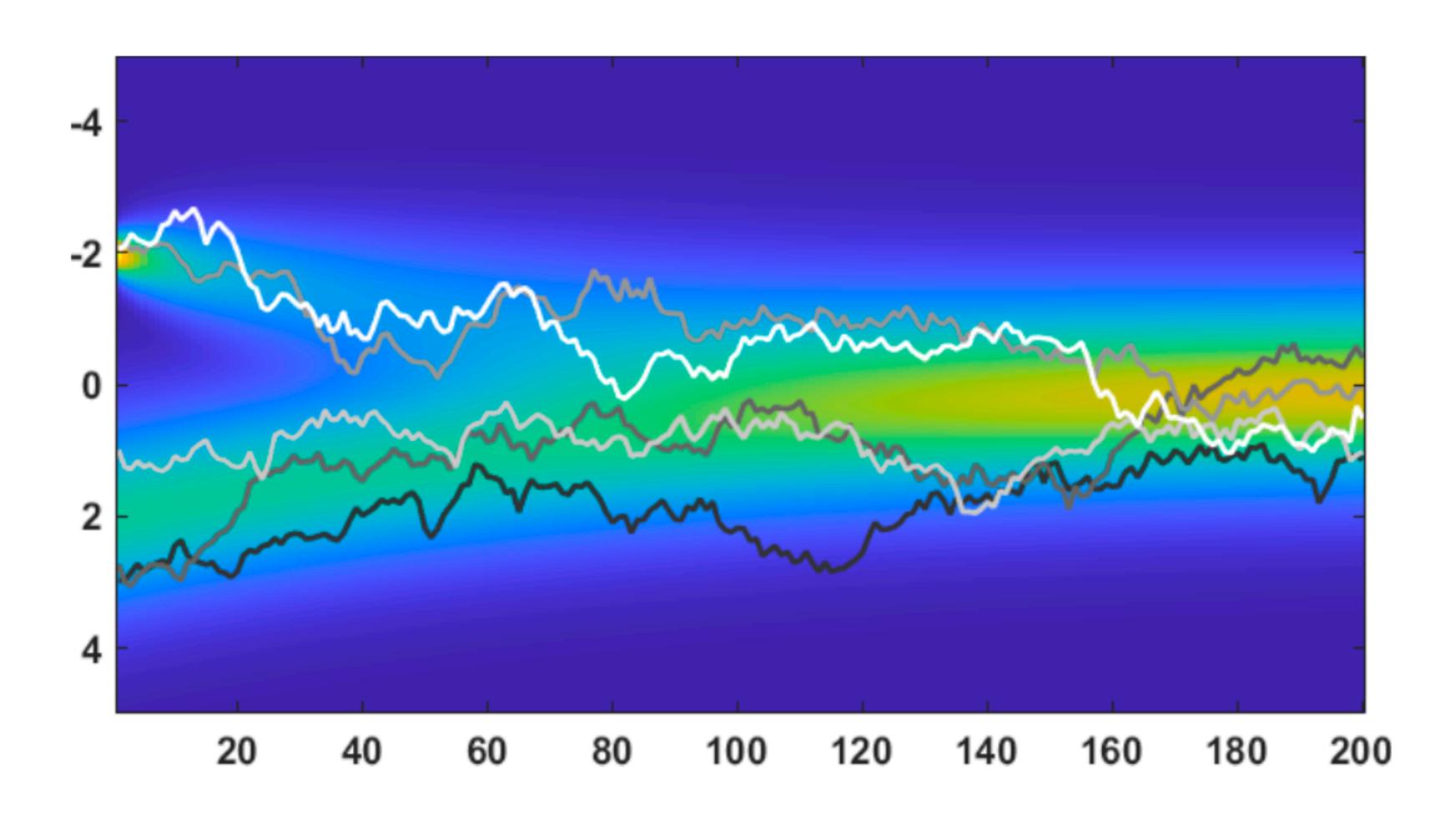


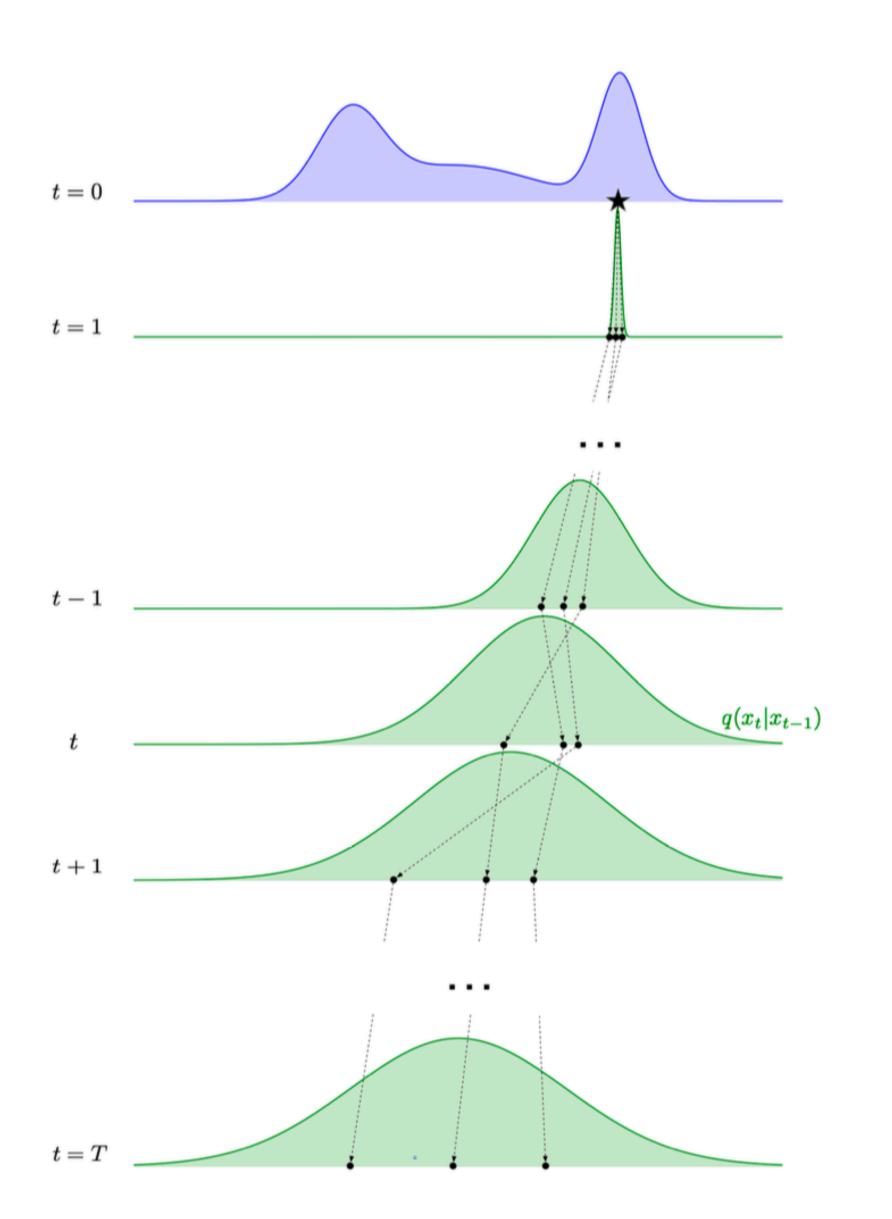
An image generated with Stable Diffusion 3.5 based on the text prompt a photograph of an astronaut riding a horse

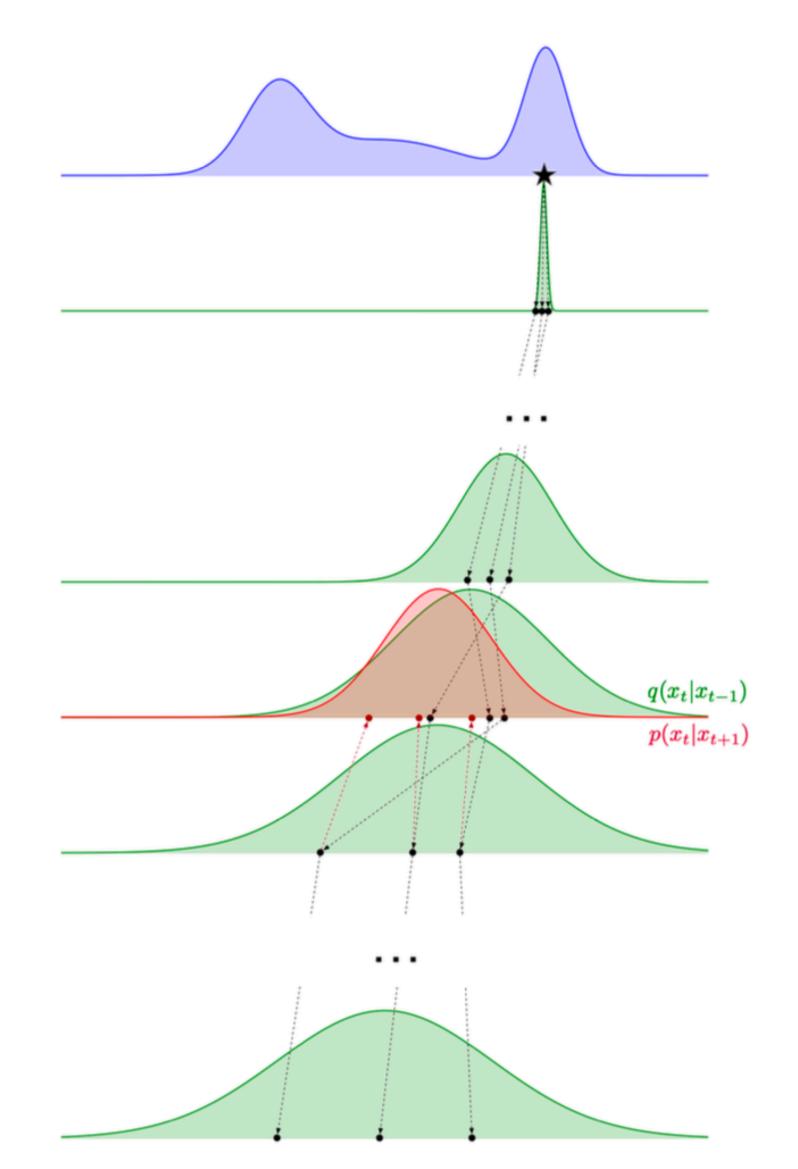




Forward Evolution







Summary and Outlook

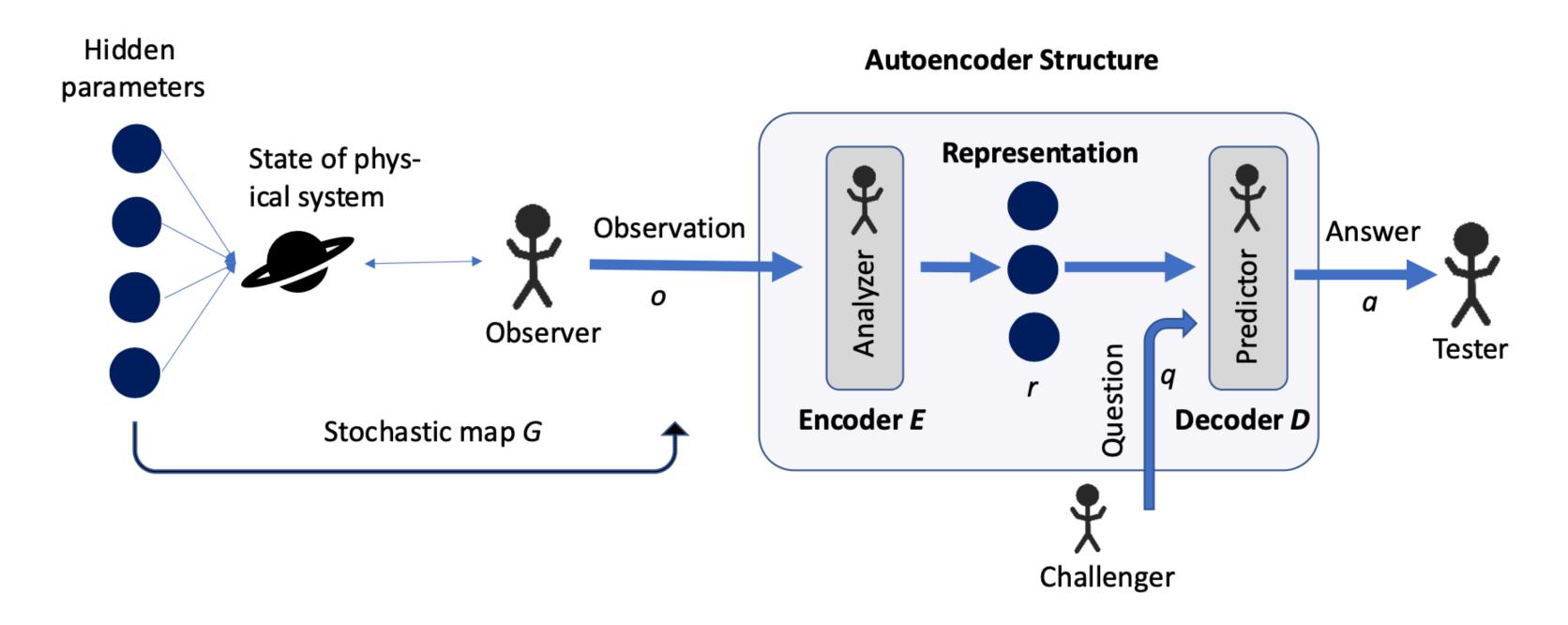


FIG. 7. Adapted autoencoder structure for discovering physical parameters. (Figure taken and modified from [330]). We assume that a small set of hidden parameters fully determines the state of a physical system. The mapping G from these hidden parameters to the measurement outputs of an observing agent is allowed to be stochastic. An autoencoder structure can then be employed to recover the relevant hidden parameters. Specifically, the observing agent provides the observation o to an analyzing agent (described by an encoding map E), whose task is to recover the relevant hidden parameters r necessary to respond to questions posed by a challenging agent. A predicting agent, which answers a question q based on the representation r, is modeled by a decoder mapping p. Finally, a testing agent compares the answer p given by the predicting agent with the correct answer obtained through direct measurements of the environment. Note that, in practice, the roles of the analyzer, predictor, and challenger could be performed by a single physicist.

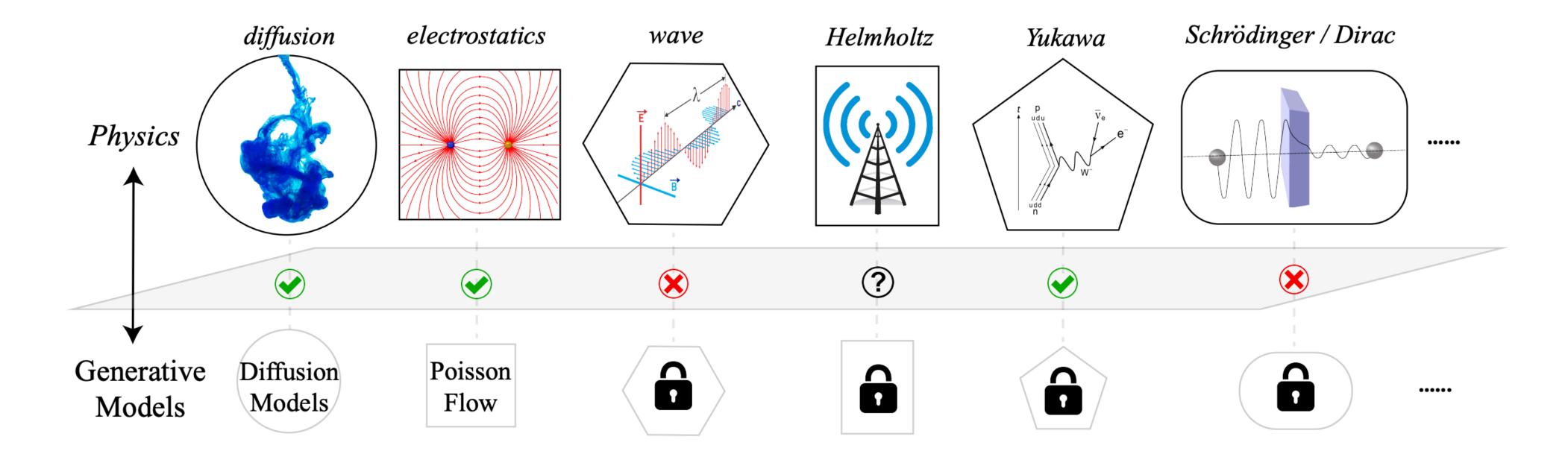


Figure 1: Duality between physics and generative models. So far only diffusion models and Poisson flow models are discovered in the literature. Can we unlock more?

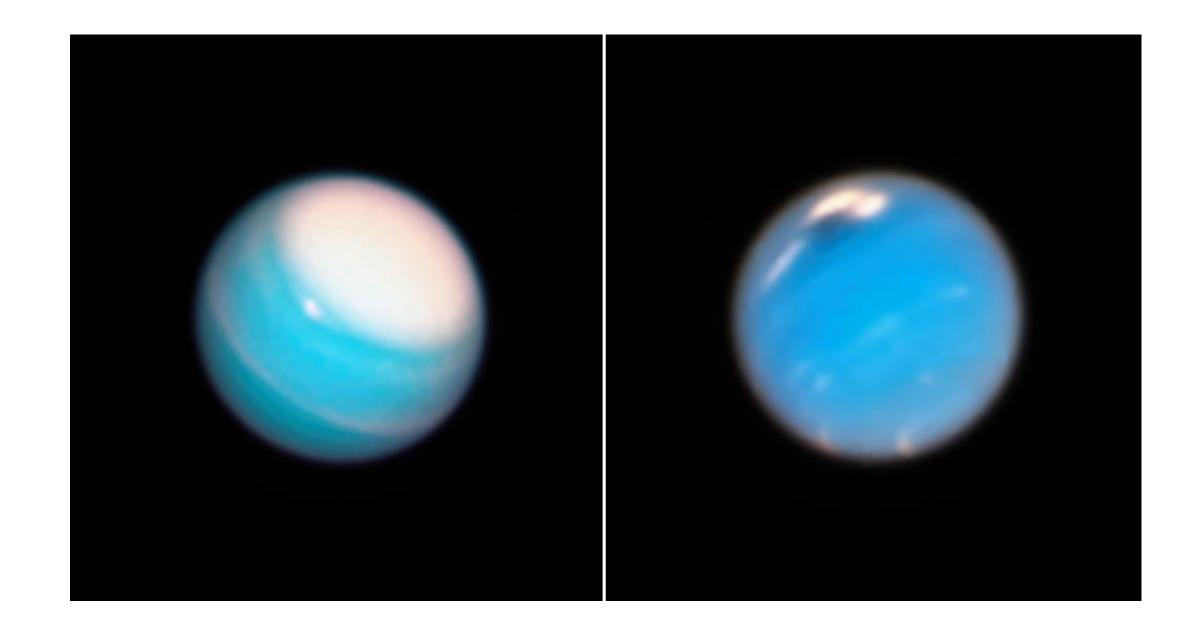
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Outlook

Nature is a teaching machine. We just begin to learn Al from it.

Back-up

Discovery of Neptune (海王星)



• In 1846, astronomers discovered Neptune. The discovery was made based on mathematical calculations of its predicted position due to observed perturbations in the orbit of the planet Uranus (天王星).