

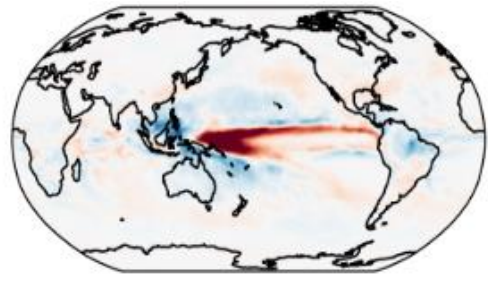
Reassessing the Scaling of **AI-Powered** vs **Dynamical** Climate Models

Upcoming survey: AI Pathways from Weather to Climate

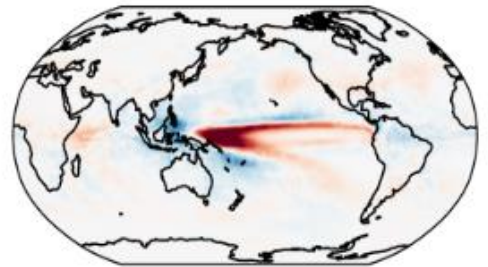
Tom Beucler (UNIL); David Neelin (UCLA); Hui Su (HKUST),
Christopher Bretherton, Oliver Watt-Meyer (AI2); Will Chapman (UC Boulder);
Costa Christopoulos, Tapio Schneider (Caltech), Ignacio Lopez-Gomez (Google),
Aditya Grover (UCLA); Adam Subel, Laure Zanna (NYU)

AI-Powered climate models are already successful...

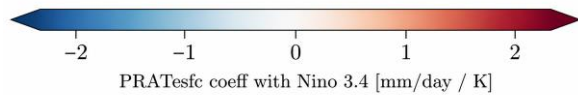
Samudra+ACE2



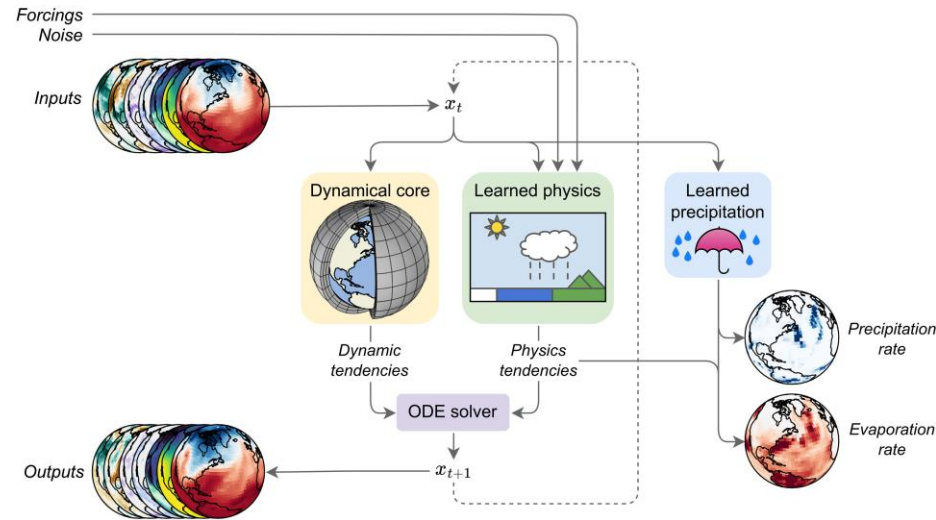
Generated



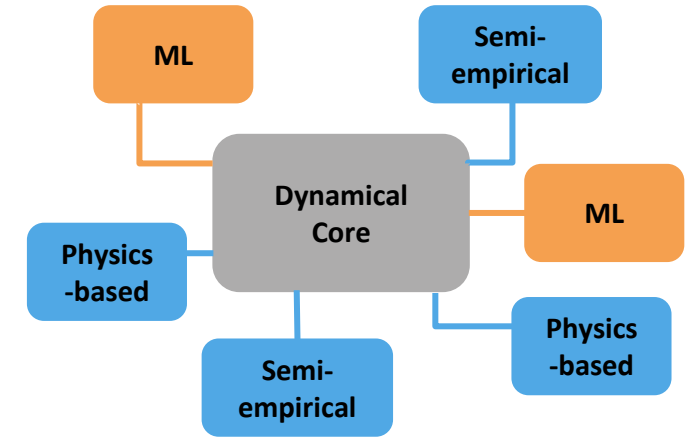
Target



Neural GCM

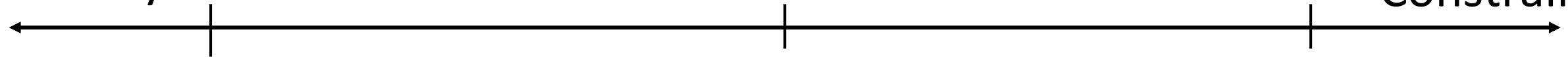


ICON-A-MLe



Physical Constraints

Flexibility



...but are **AI-powered** climate models intrinsically more efficient than **dynamical** ones?

1) Earth system science perspective:

- Cheaper models → more diverse set of experiments (e.g., scenarios)
- Larger ensembles → sharper uncertainty estimation

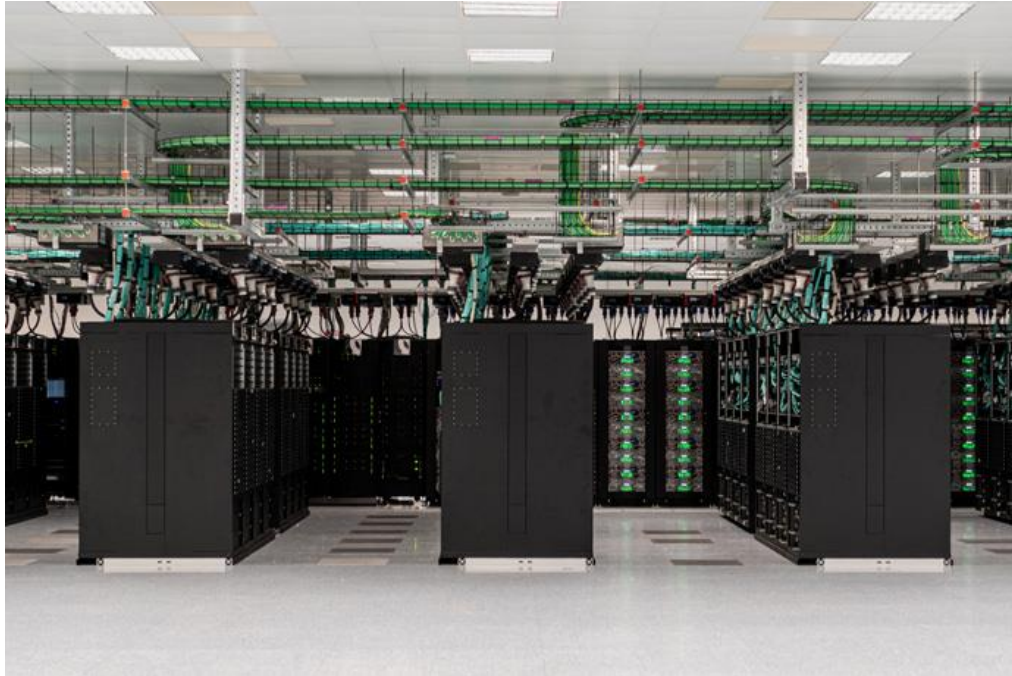
2) Faster model development cycles accelerate progress (e.g., AI for weather)

3) Sustainability perspective:

Larger added scientific value for a given negative environmental impact

...but are **AI-powered** climate models intrinsically more efficient than **dynamical** ones?

CPU-Based



GPU/TPU-Based



≠



Image sources: ECMWF Supercomputer facility, NVIDIA A100, Google Cloud Blog (TPU v4 pod)

...but are **AI-powered** climate models intrinsically more efficient than **dyn.** ones? Confounders include:

- ⚠️ CPU- vs. GPU/TPU-based simulation
- ⚠️ Spatial vs. spatiotemporal resolution
- ⚠️ Hardware characteristics
- ⚠️ Strong vs. weak scaling
- ⚠️ Simulation's *effective* resolution
- ⚠️ Simulation's agreement with reference data ("skill") & robustness

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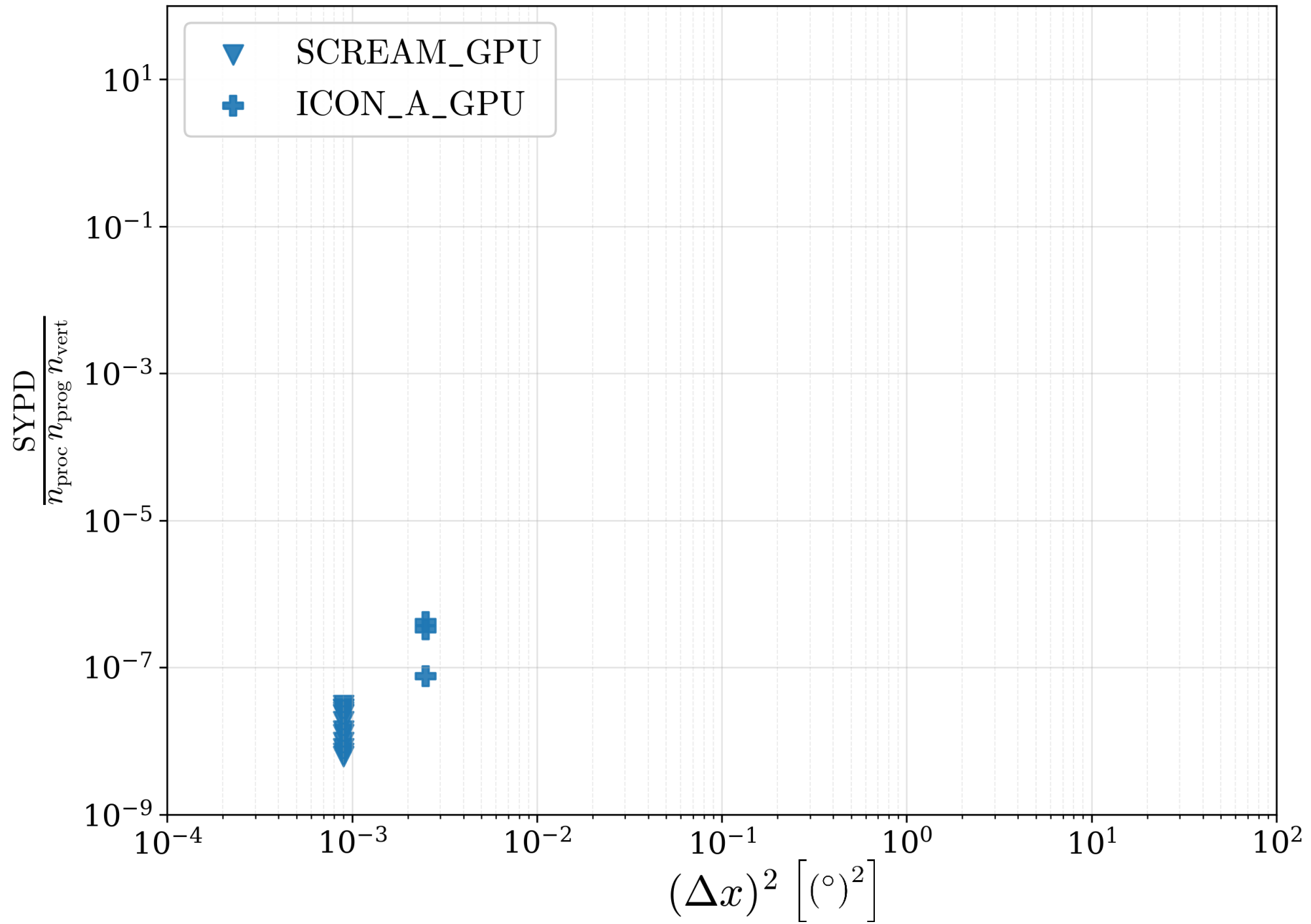


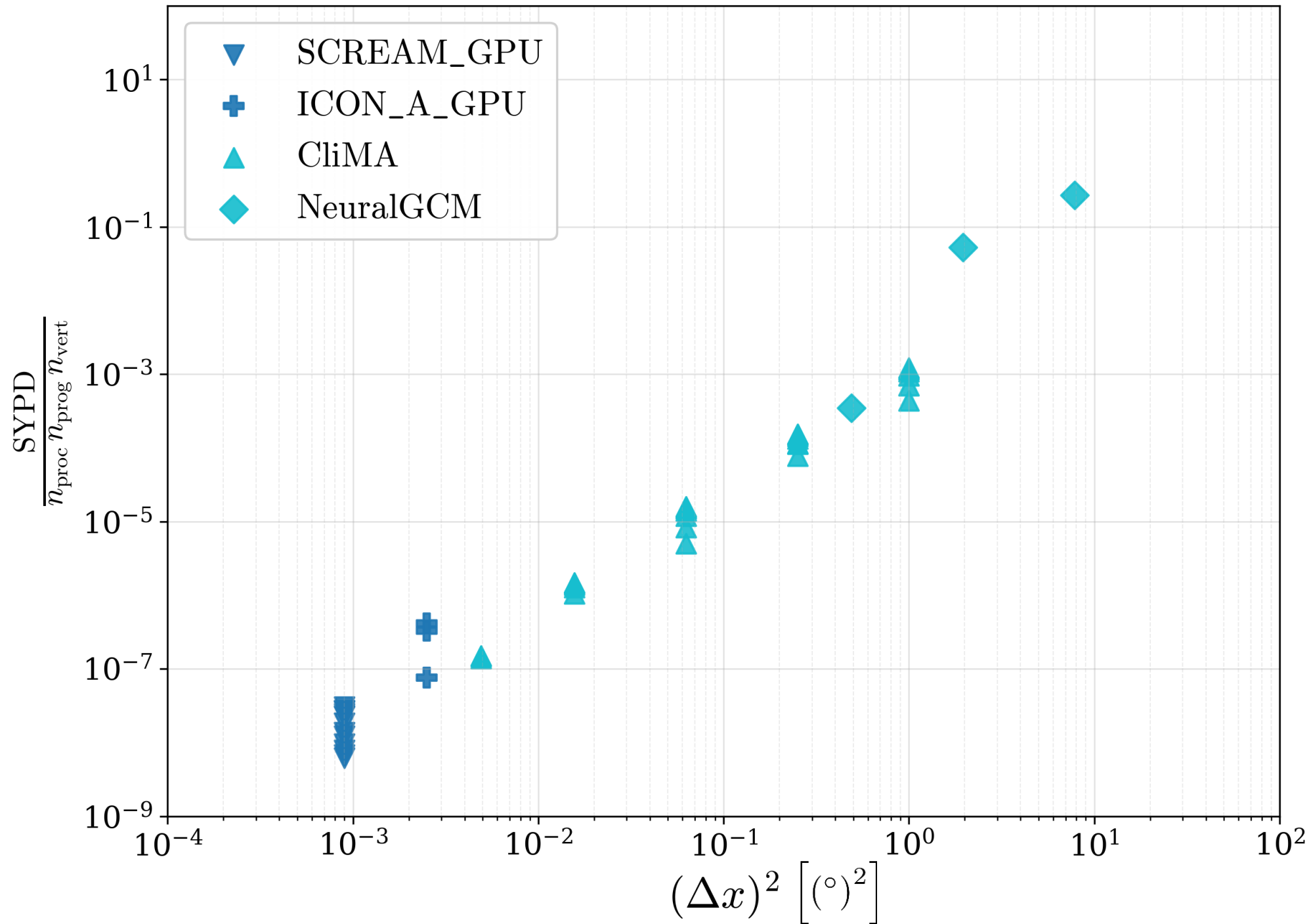
Image source: M. Johnson (Wikimedia Commons)

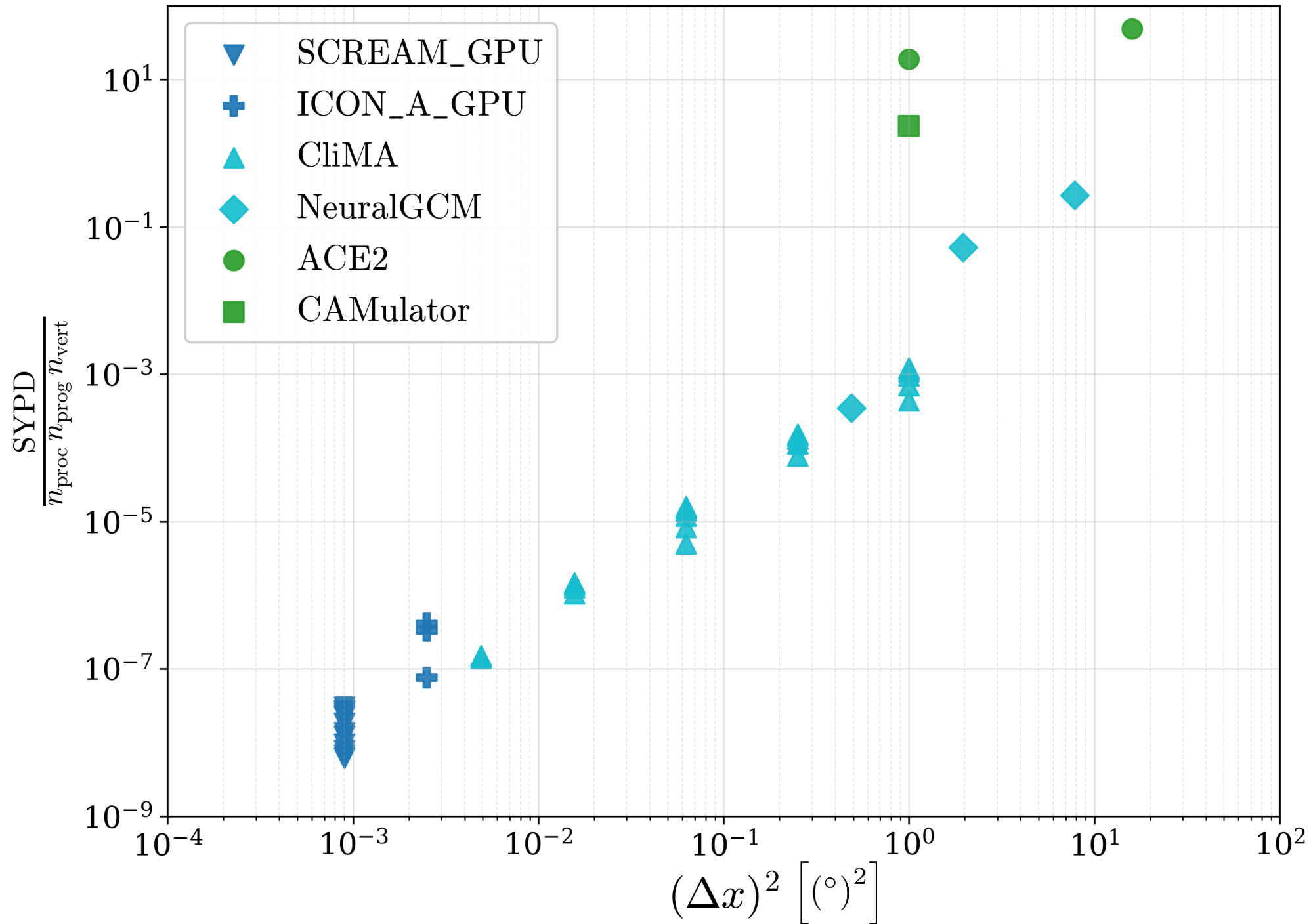
When considering raw throughput for a given spatial resolution, **AI models** appear more efficient...

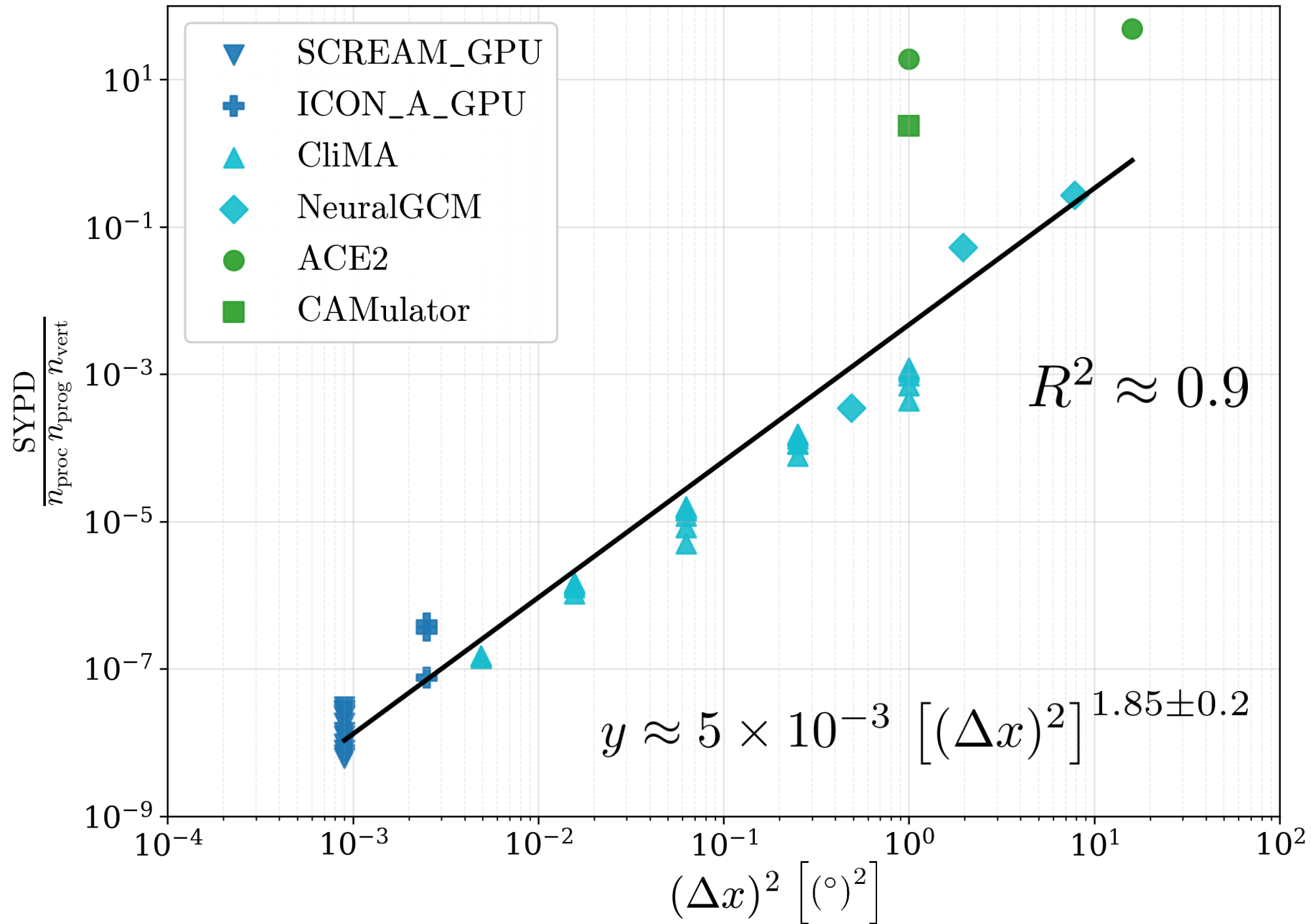
$$\frac{\text{SYPD}}{n_{\text{proc}}} \sim n_{\text{ens}} n_{\text{prog}} (\Delta x) (\Delta y) (\Delta z)$$

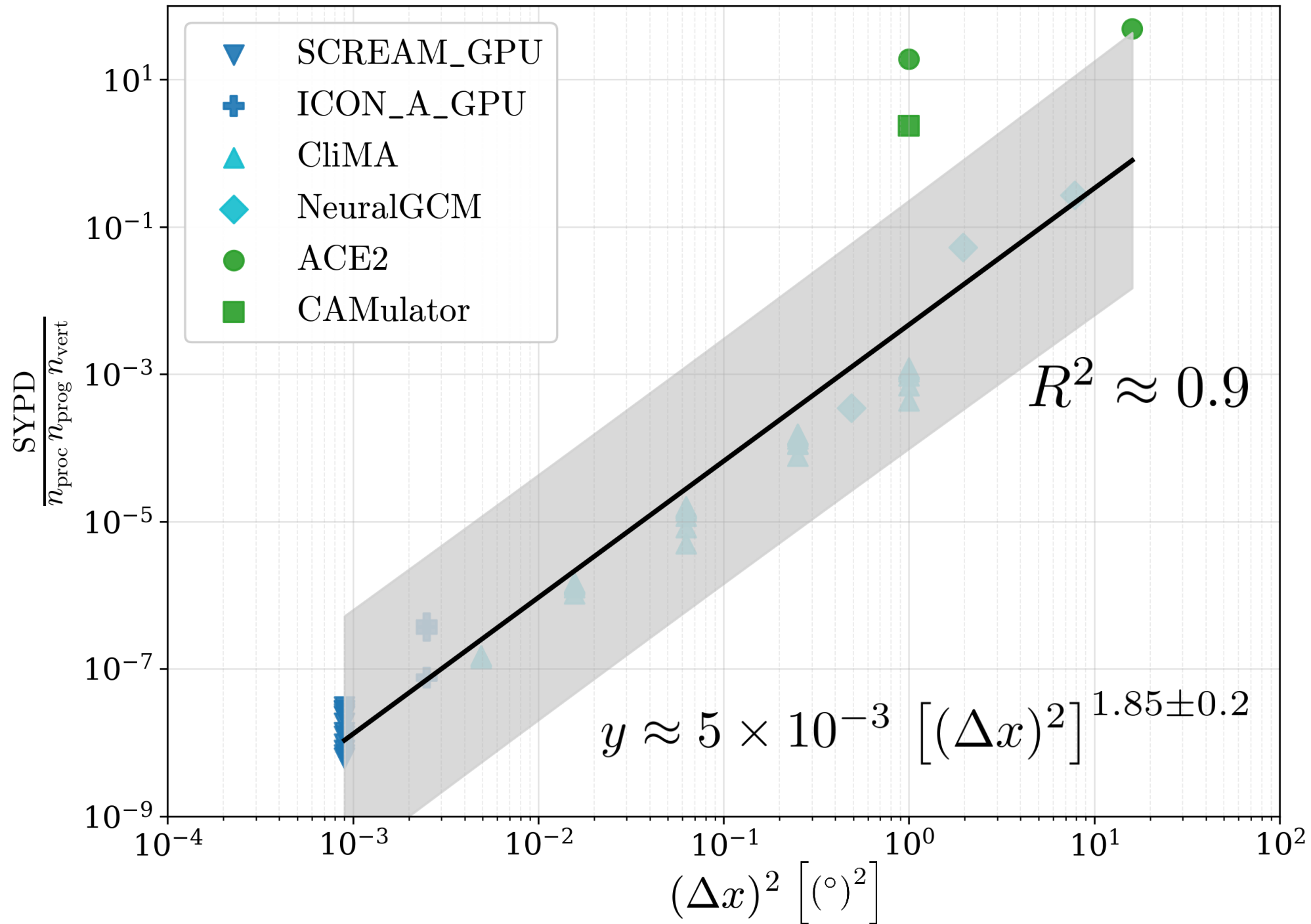
$$\frac{\text{SYPD}}{n_{\text{proc}} n_{\text{prog}} n_{\text{vert}}} \text{ vs. } (\Delta x)^2$$











When considering raw throughput for a given spatial resolution, **AI models** appear more efficient...

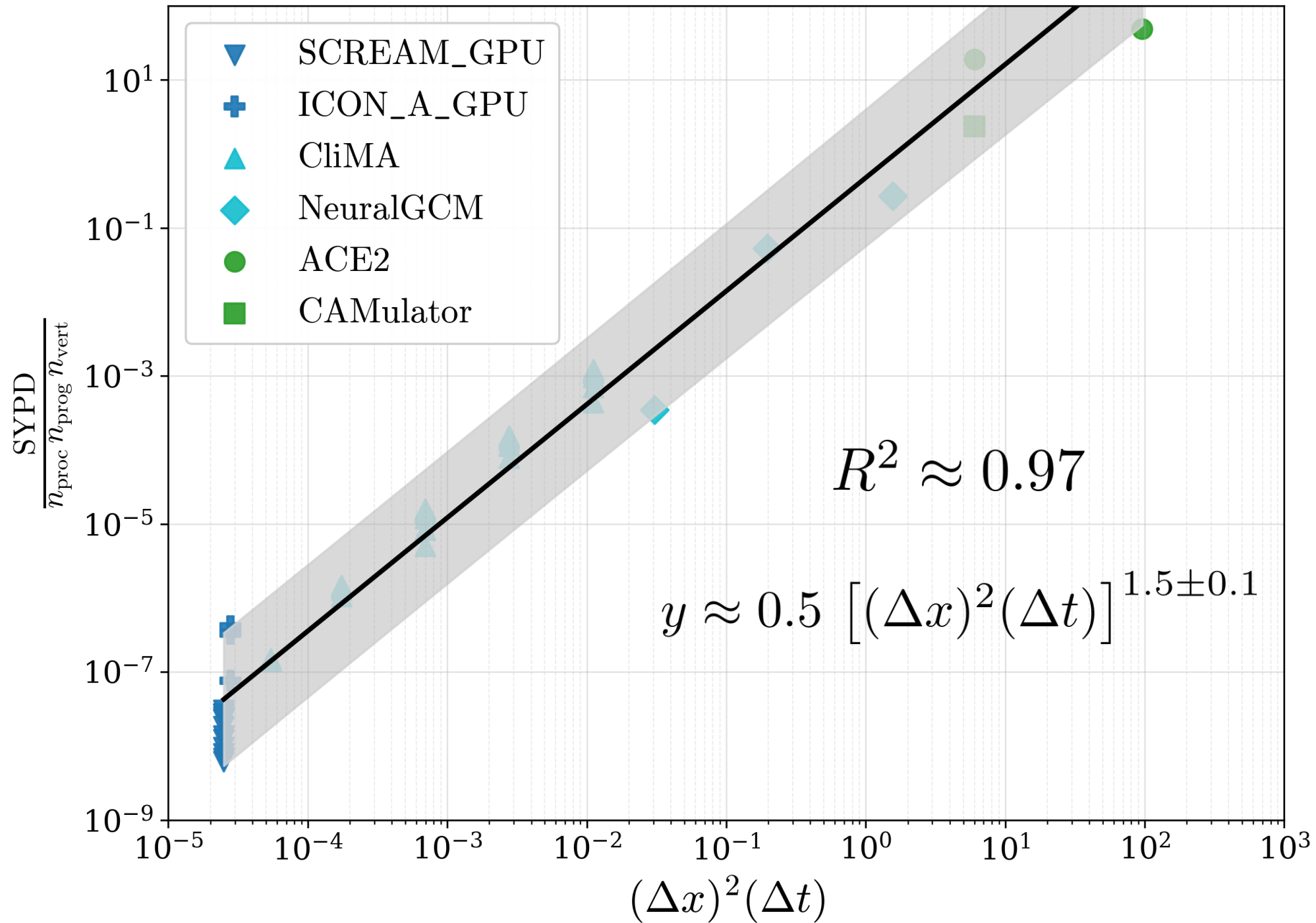
✓ CPU- vs. GPU/TPU-based simulation



Image source: M. Johnson (Wikimedia Commons)

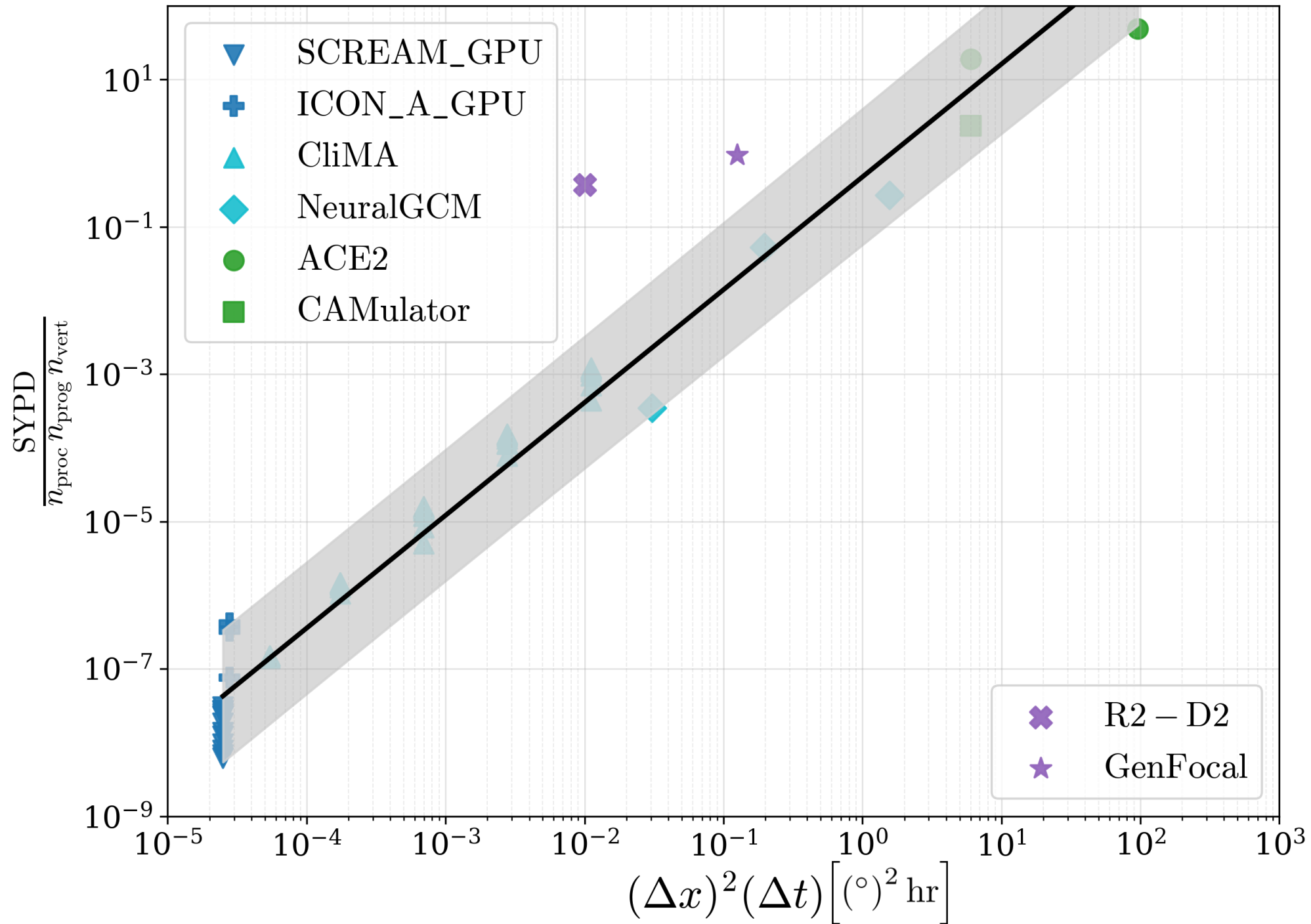
...but once the timestep is accounted for, the improved efficiency of **AI models** is less clear

$$\frac{\text{SYPD}}{n_{\text{proc}} n_{\text{prog}} n_{\text{vert}}} \text{ vs. } (\Delta x)^2 (\Delta t)$$



Are **AI-powered** climate models intrinsically more efficient than **dynamical** ones? Confounders include:

- ✓ CPU- vs. GPU/TPU-based simulation
- ✓ Spatial vs. spatiotemporal resolution
- ⚠ Hardware characteristics
- ⚠ Strong vs. weak scaling
- ⚠ Simulation's agreement with reference data ("skill") & robustness
- ⚠ Simulation's *effective* resolution



Reported throughput/resolution challenge the hypothesis that AI-powered climate model are intrinsically more efficient than their GPU dynamical counterparts but...

1. AI climate model emulators allow **task-targeted prediction**
2. This avoids **integrating the full high-frequency, multivariate state at the short time step traditionally required for numerical stability**
3. **Downscaling/post-processing** can substitute explicit fine-scale resolution when observations are available for inexpensive local hazard assessment



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