Neural Ideal Large Eddy Simulation: Modeling Turbulence with Neural Stochastic Differential Equations

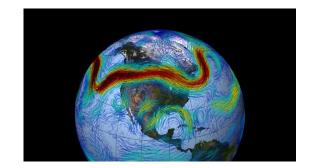
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Closure modeling: motivation

- Chaotic physical flows are **expensive** to simulate accurately
 - Since they require resolving **multiple scales** spanning several orders of magnitude
- Practitioners resort to **coarse-grained** simulations
 - However they veer off-course since don't account for the effect of small scales on the large scales



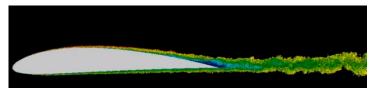


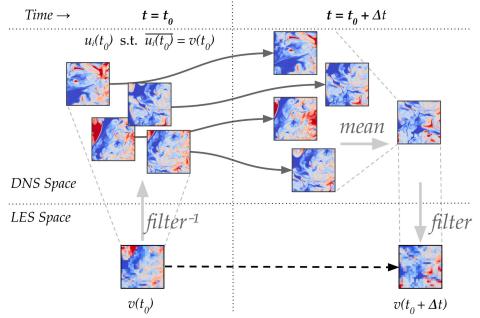
Image courtesy: NASA, Nek5000



Ideal Large Eddy Simulation (LES)

- An LES, or reduced order, field v could represent any full-order field \overline{u} such that $\overline{u} = v$
- Ideal LES (Langford and Moser 1999) takes the conditional average of *all possible* small scale effects consistent with the large scale state v
- Mathematically,

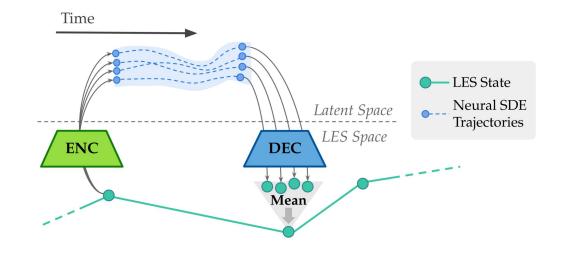
$$rac{\partial v}{\partial t} = \mathbb{E}_{\pi_t} \left[\left. rac{\overline{\partial u}}{\partial t} \right| \overline{u} = v
ight]$$



Can we make architectural choices which bake in the inductive bias of Ideal LES?

Neural ideal LES (niLES)

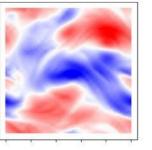
- Given LES states $\overline{u(t_0)}$ and $\overline{u(t_1)}$, model distribution of full-order trajectories on $[t_0, t_1]$
- Learn a data-driven neural SDE parameterized by a neural network evolving on a fast timescale τ
- SDE evolves a latent state much smaller dimensionality
- With enough data, SDE should discover the underlying solution manifold

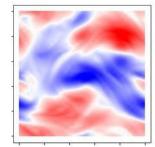


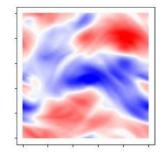


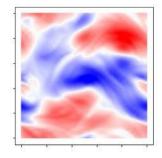


Velocity (x direction)

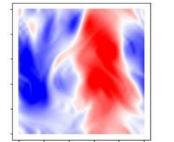




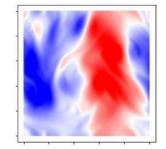


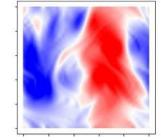


Velocity (y direction)

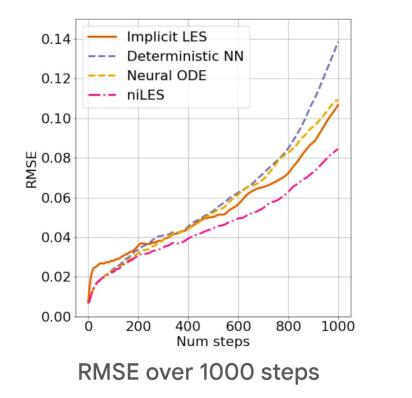


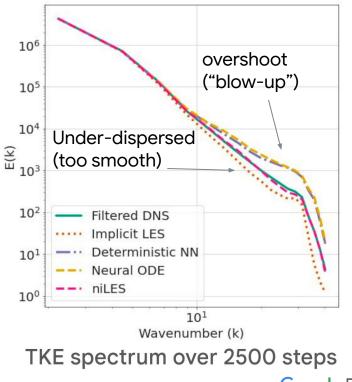






Accuracy and long-term statistics





Thank You

