

Stochastic dynamical wake modeling for wind farms

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RESEARCH PROBLEM

Low-fidelity wake modeling

The oversimplified static nature of low-fidelity wake models that do not account for the complex aerodynamic interactions among turbines limits their utility for closed-loop model-based wind farm control.

Question

Can we use data to introduce dynamical augmentations to low-fidelity static wake models and accurately capture power and thrust force measurements in accordance with large-eddy simulations (LES) and SCADA data?

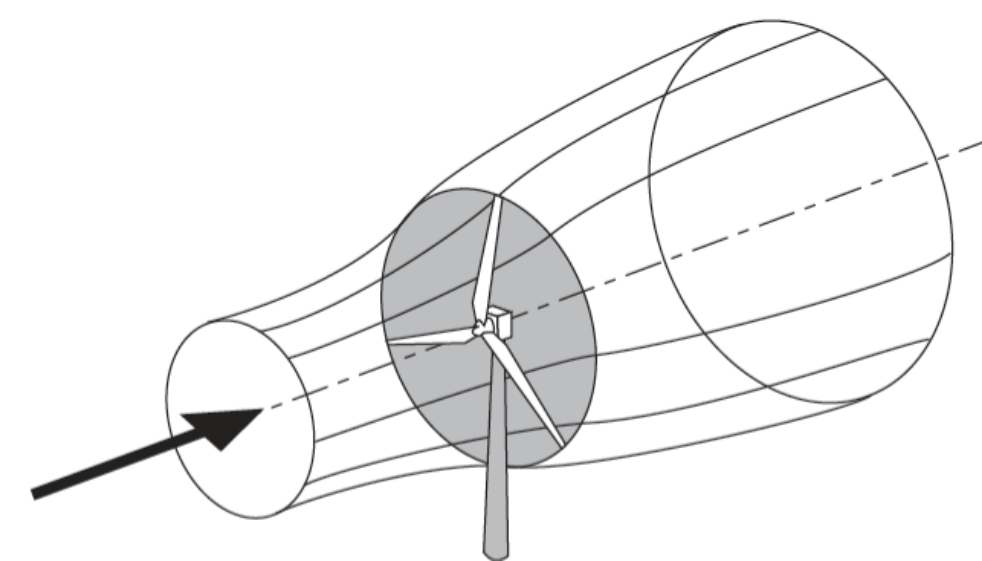
LOW-FIDELITY WAKE MODELING

The actuator disk concept for energy extraction:

The turbine rotor is modeled as a disk and its behaviour is analysed within a control volume.

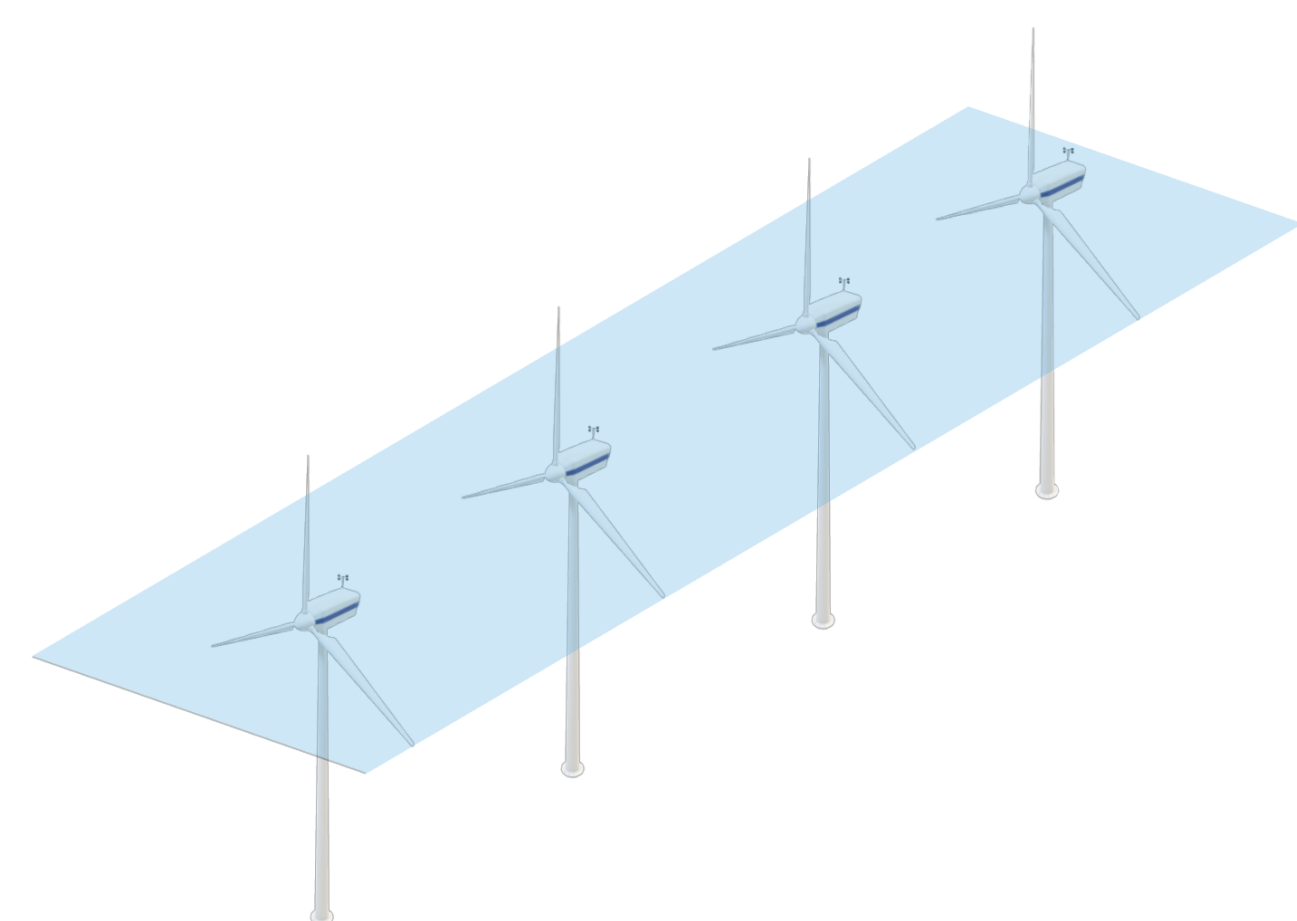
$$\text{Thrust force: } F = \frac{1}{2} \rho A C_T u^2$$

$$\text{Extracted power: } P = \frac{1}{2} \rho A C_P u^3$$

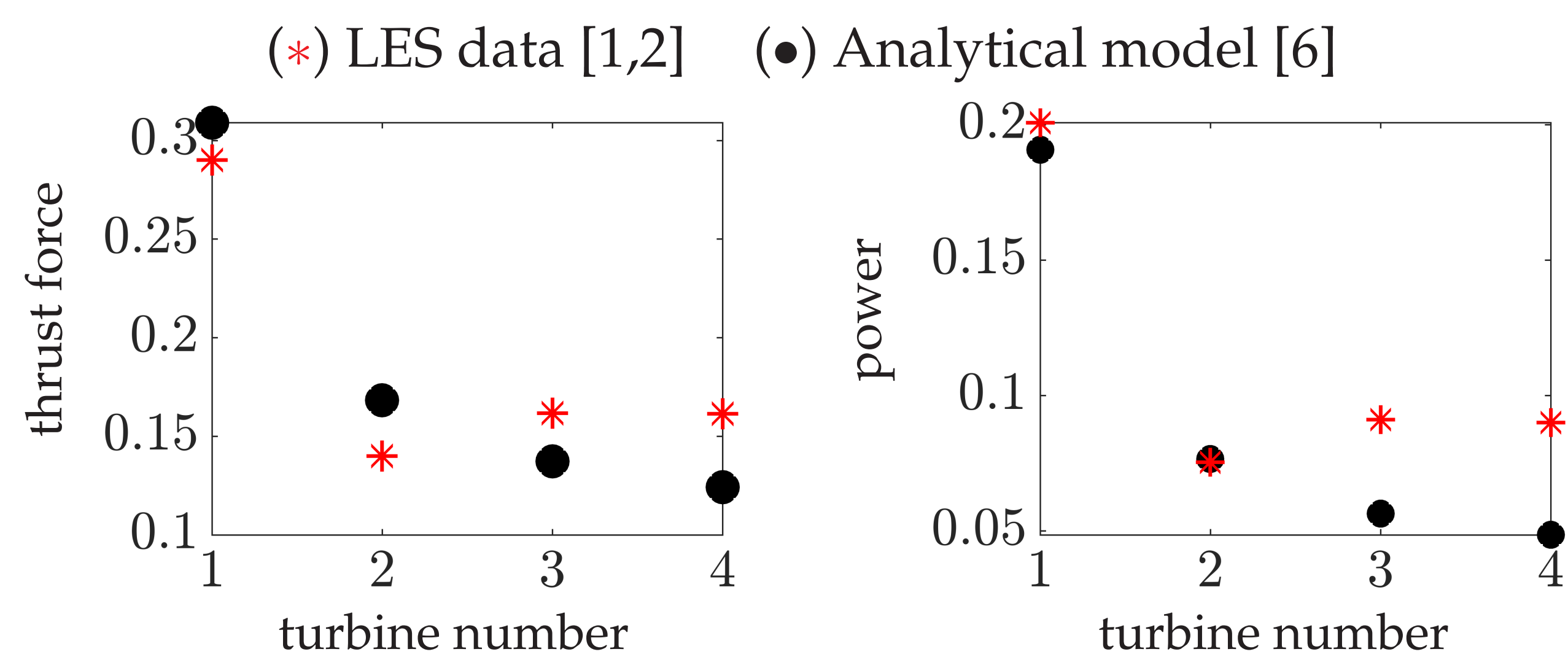


Analytical wake models:

- Jensen model
- Park model
- Frandsen model
- Gaussian deficit model

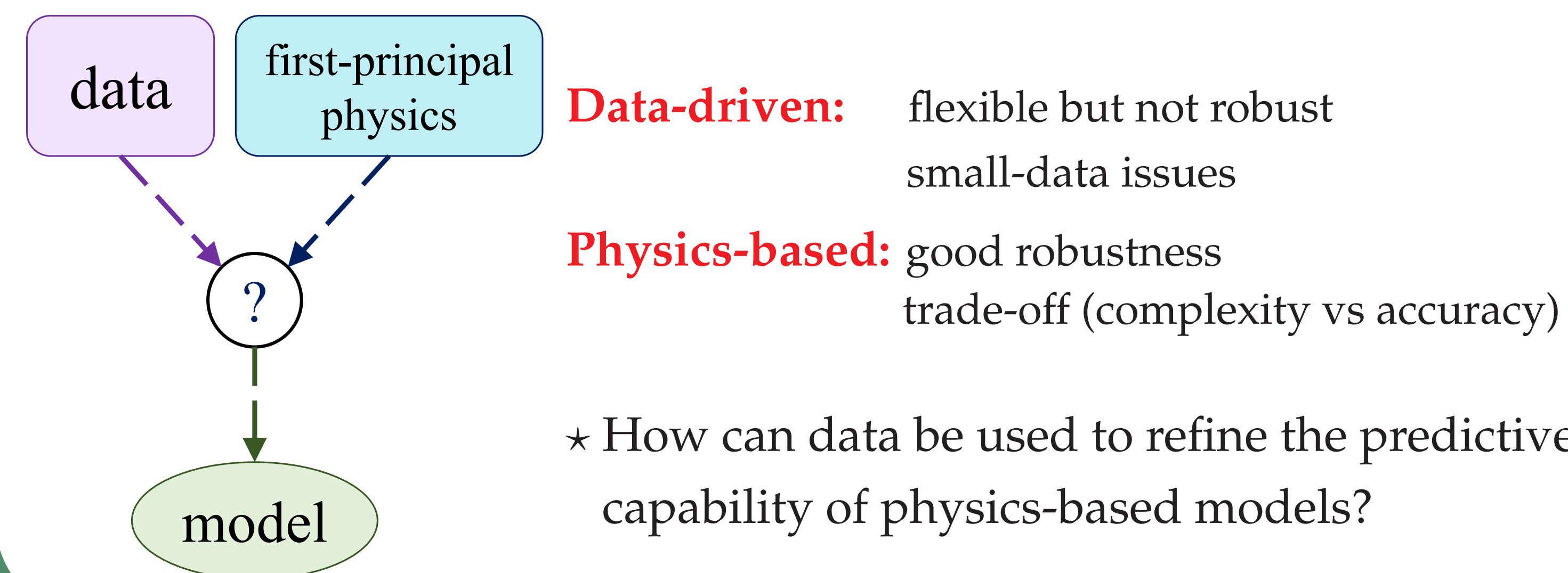


★ Example:
4 × 1 wind farm



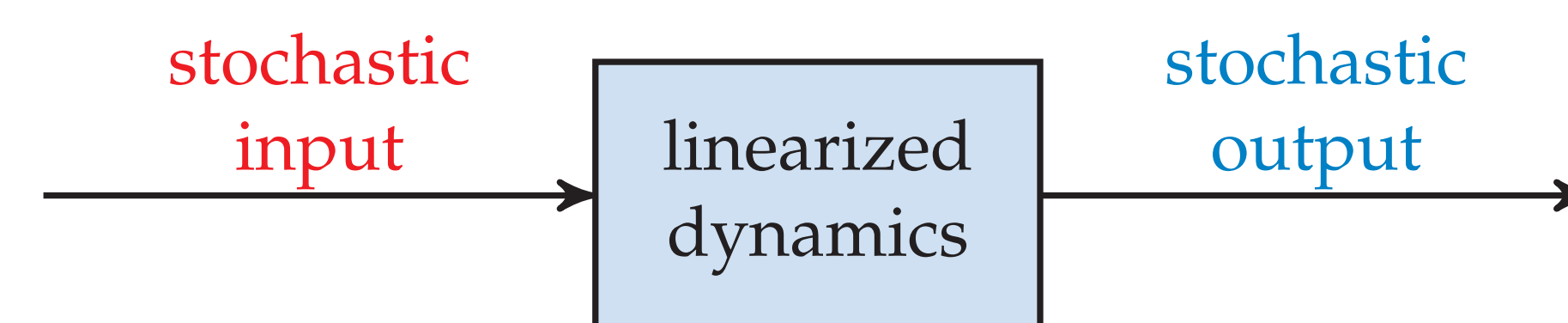
★ Model predictions fail to capture **quantities** and **trends**.

DATA-DRIVEN vs PHYSICS-BASED MODELING



APPROACH: STOCHASTIC DYNAMICAL MODELING [3,4]

Stochastically forced linearized equations



- view **second-order statistics** as **data** for **inverse problems**
- identify **input statistics** to account for **available velocity statistics** in output

STOCHASTIC WAKE MODELS [5]

- Model dynamics of fluctuations (\mathbf{v}) around base flow ($\bar{\mathbf{u}}$):

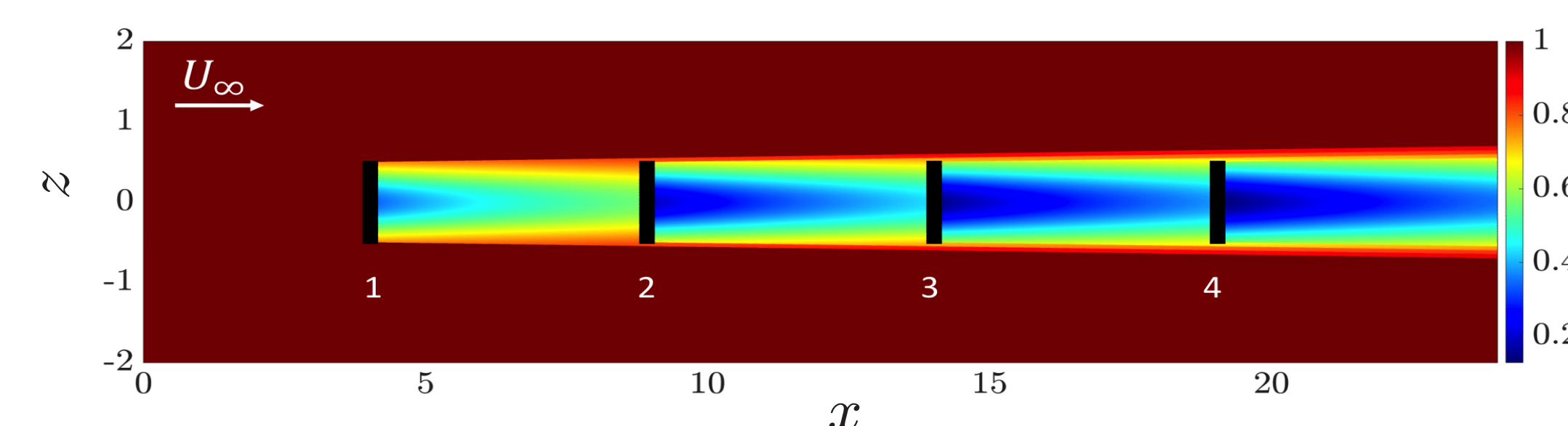
$$\mathbf{u} = \bar{\mathbf{u}} + \mathbf{v} \quad \bar{\mathbf{u}} = [\mathbf{u}] \quad [\mathbf{v}] = 0$$

$$F = \frac{1}{2} \rho A C_T (\bar{\mathbf{u}}^2 + \overline{\mathbf{v}^2}) \quad P = \frac{1}{2} \rho A C_P (\bar{\mathbf{u}}^3 + 3 \bar{\mathbf{u}} \overline{\mathbf{v}^2})$$

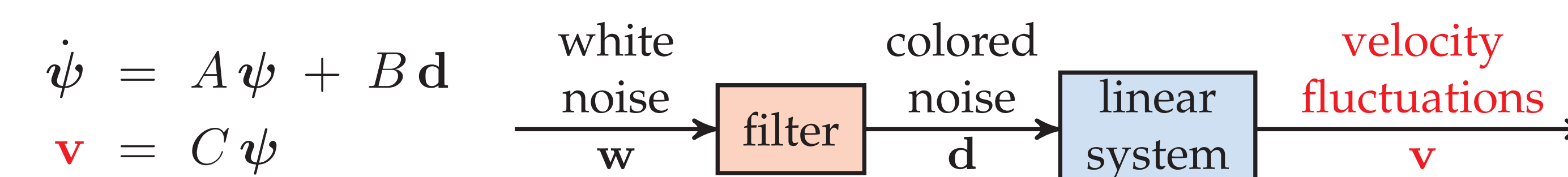
$$\overline{\mathbf{v}^2} := \int_S [\mathbf{v}(\mathbf{x}, t)^2] dx$$

- ★ Analytical base flow $\bar{\mathbf{u}}$: Gaussian deficit model [6]

$$\bar{\mathbf{u}}(x, z) = U_\infty - U_\infty \left(1 - \sqrt{1 - \frac{C_T}{8 (k^* x / d_0 + 0.2 \sqrt{\beta})^2}} \right) e^{-\frac{1}{2 (k^* x / d_0 + 0.2 \sqrt{\beta})^2} \left(\frac{z}{d_0} \right)^2}$$

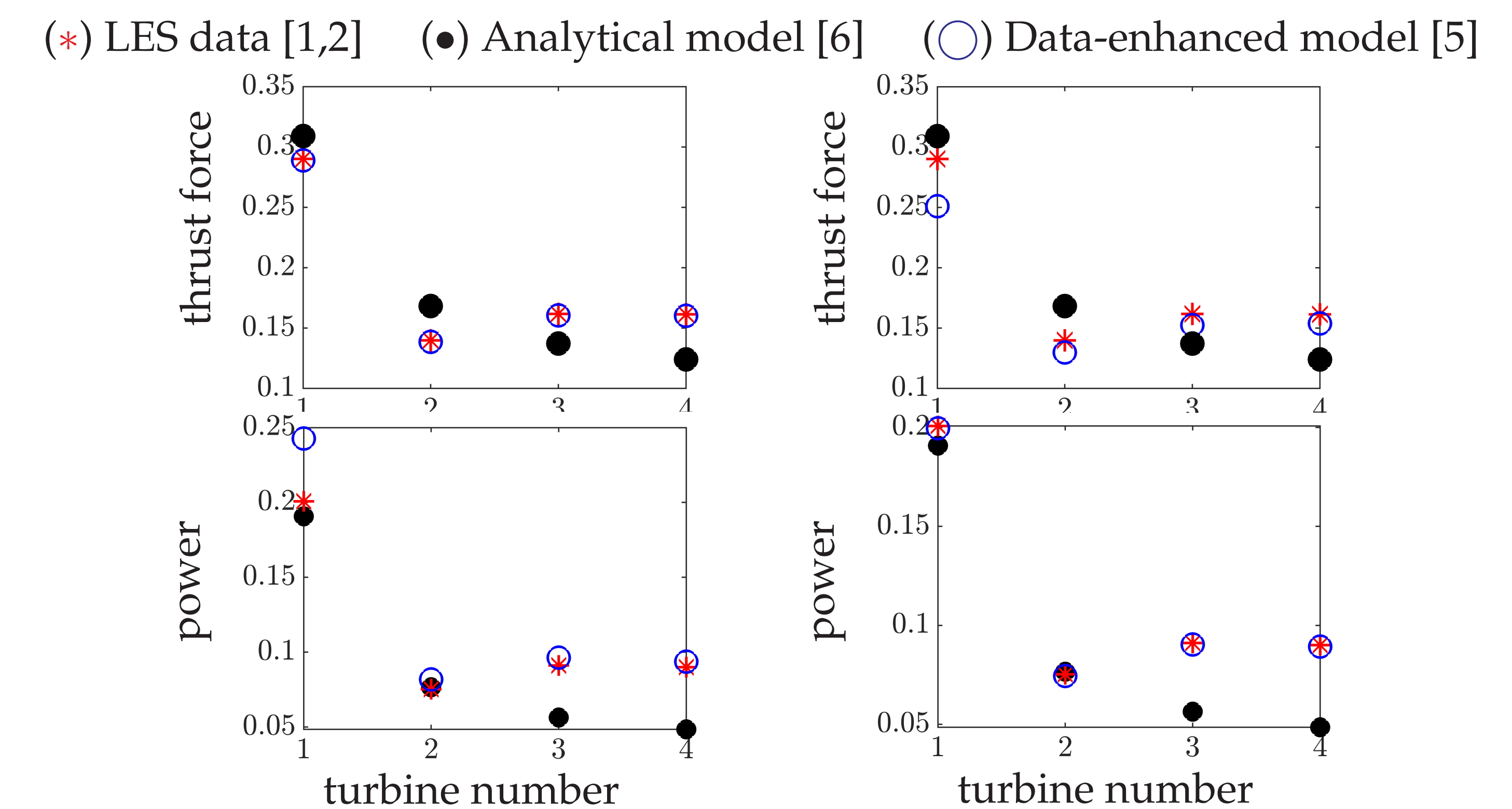


- ★ Model fluctuation dynamics using linearized Navier-Stokes and shape forcing statistics to match LES data [5]

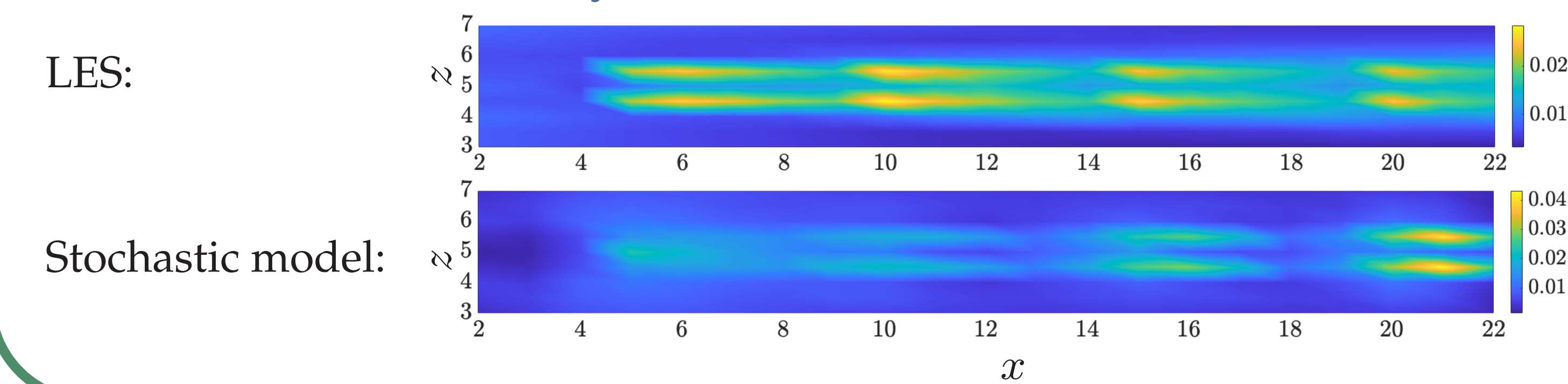


RESULTS

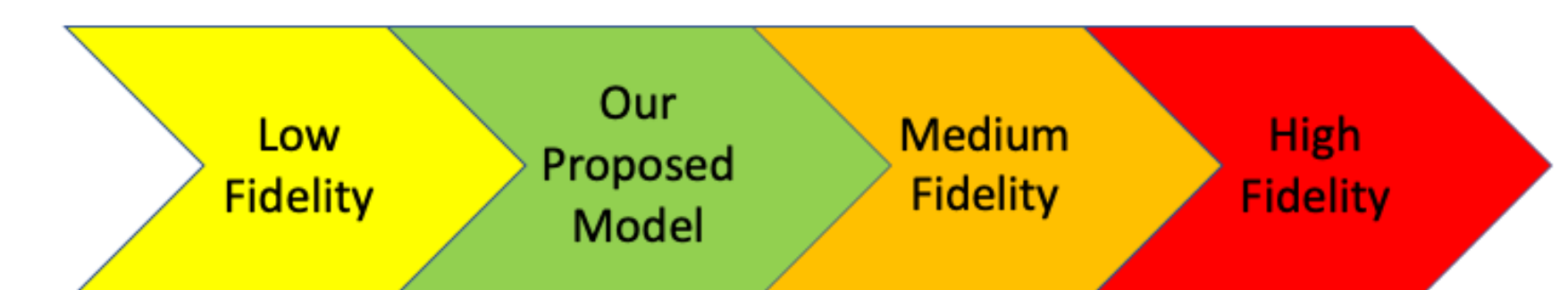
- Predictions of power and thrust force



- Prediction of velocity intensities



CONCLUSION



Stochastic dynamical models improve predictive capability of low-fidelity wake models in capturing power and thrust force in wind farms, in addition to turbulent intensities.

Outlook:

- Extension to 3D flows to capture vortex shedding effects and wake curl
- Model-based feedback control

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- [6] M. Bastankhah and F. Porté-Agel, "A new analytical model for wind-turbine wakes", *Renewable Energy*, vol. 70, pp. 116-123, 2014.