

WindMil: Equivariant Graph Learning for Wind Loading Prediction

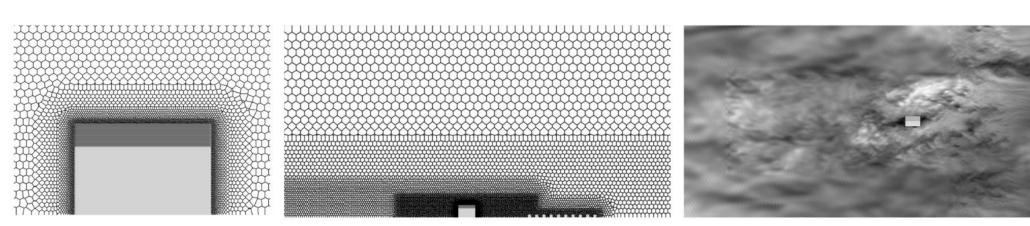
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Motivation

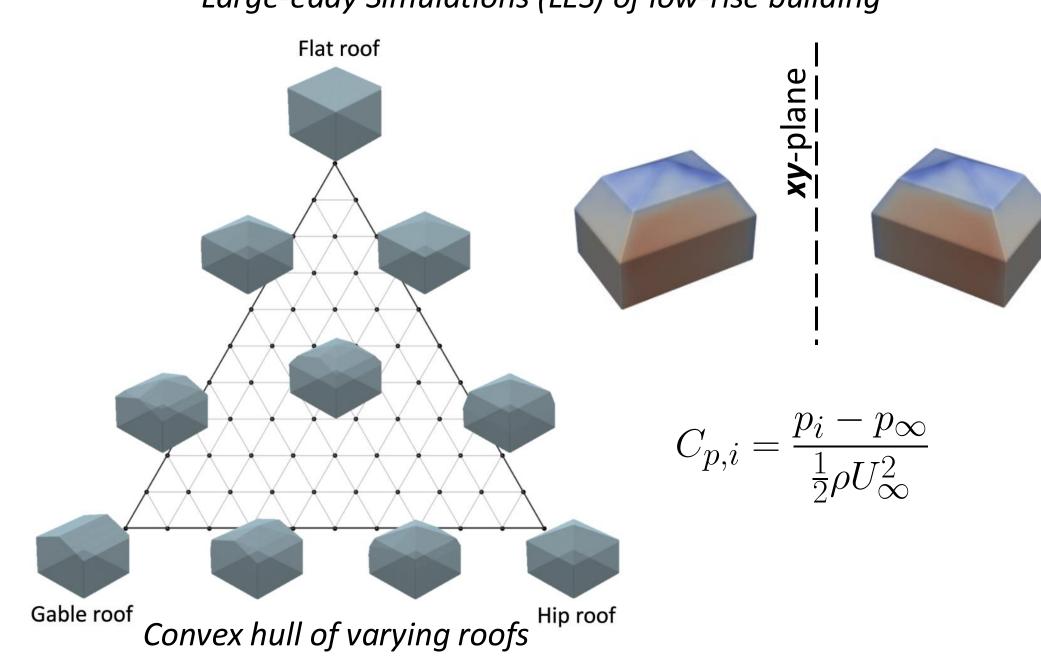
Accurately predicting wind loads on buildings is crucial for safety and sustainable design. Wind tunnels and Computational Fluid Dynamics are accurate but prohibitively expensive for engineering use; we propose **WindMiL,** a **reflection-equivariant GNN surrogate** for fast, accurate and physically consistent predictions.

Contributions

- Dataset: Large-scale dataset of wind loading on low-rise buildings by systematically varying roof morphologies and wind directions using LES.
- **Model**: Reflection-equivariant GNN that respects reflectional invariances to *xy*-plane

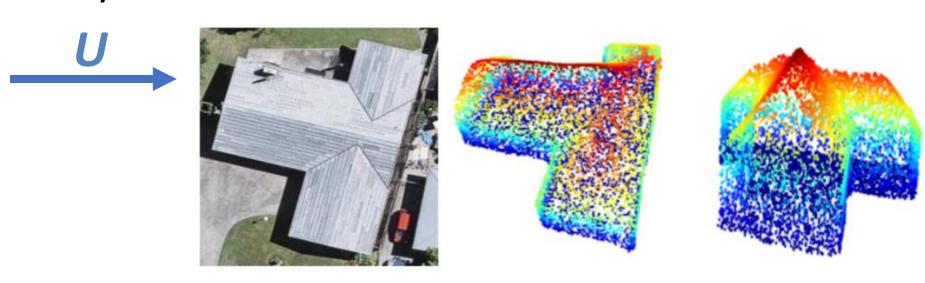


Large-eddy Simulations (LES) of low-rise building

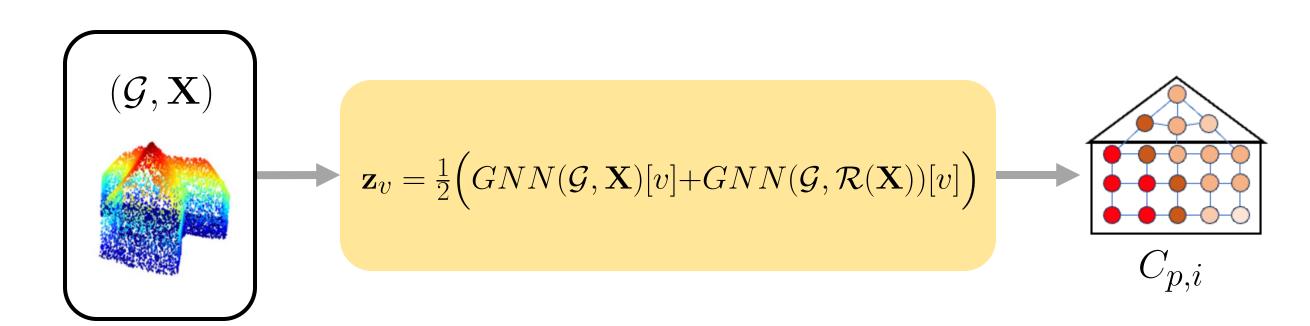


Graph representation & Model Architecture

Graph: Represent buildings as point-clouds where nodes are surface vertices with features $\mathbf{x}_v = (x, y, z, n_x, n_y, n_z)$; edges represent mesh connectivity.



WindMiL equivariant model: Define $\mathcal{R}:(z,n_z)\mapsto (-z,-n_z)$ Graph structure and geometrical features of original and reflected building pass through the **same GNN** and resulting embeddings are averaged. Then, feed-forward predictor outputs the node-level $C_{p,i}$.



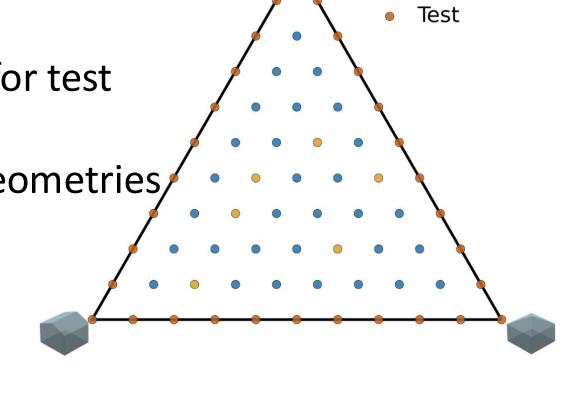
Evaluation Method

Interpolation: random train/dev/test inside convex hull

Extrapolation: hold out boundary for test

+Sym: test set includes reflected geometries/

Metrics: MAE and hitrate (+/- 10%)



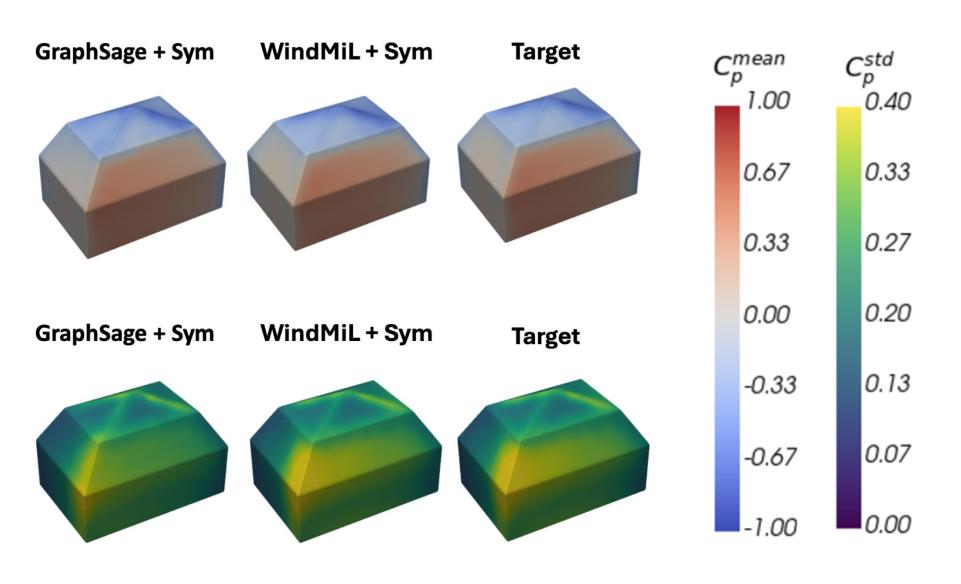
Train

Validation

Results

 WindMiL outperforms non-equivariant GNN on extrapolation and +Sym tests, showing higher generalization.

Method	MAE	Hitrate (%)
GRAPHSAGE (Mean C_p)	0.033	94.98
GRAPHSAGE + Sym (Mean C_p)	0.054	84.67
WINDMIL (Mean C_p)	0.035	94.16
WINDMIL + Sym (Mean C_p)	0.034	94.69
GRAPHSAGE (Std C_p)	0.012	97.35
GRAPHSAGE + Sym (Std C_p)	0.020	91.57
WINDMIL (Std C_p)	0.013	97.30
WINDMIL + Sym (Std C_p)	0.012	97.33



Takeaways

- WindMiL enforces reflection equivariance and outperforms non-equivariant GNN.
- Significantly reduces computational cost to seconds, compared to LES, which requires 24+ hours.