

# Fast urban flow predictions through Convolutional Neural Networks.

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Having real-time and accurate numerical predictions of urban wind flow can be extremely useful for developing tools intended to improve citizens' life quality and health. However, traditional methods such as Computational Fluid Dynamics (CFD) are unsuitable for fast prediction. This work proposes using Convolutional Neural Network (CNN) trained with a newly-created vast dataset to enable fast and accurate flow predictions for any urban geometry. The dataset has been generated through high-fidelity CFD simulations of 30 different European Urban areas and 90 meteorological conditions. The geometries were selected to have a wide variety of urban flow patterns and geometrical features allowing the Neural Network (NN) to learn a representative range of urban flow conditions. Then, a CNN was trained to reproduce the urban wind flow for any urban geometry and meteorological condition. The strategy allows for predicting accurate mean wind flow in urban areas that have not been seen in training time, showing good generalization properties.

## 1 Introduction

Real-time urban wind flow predictions may help improve citizens' life quality and health since they would allow taking instantaneous countermeasures to mitigate urban pollution, for instance. CFD has been traditionally used to have a detailed insight into urban flow behavior. However, CFD is not suitable for fast prediction due to its complexity and computational cost, even using low-fidelity models such Reynolds-Averaged Navier-Stokes. Thus, alternative methods are needed.

In recent years, advances in machine learning (ML) allowed leveraging the enormous volume of data generated through CFD simulations to develop data-driven reduced models to obtain numerical predictions at a reasonable time, cost, and effort (1). Urban flows have not been an exception.

In particular, generative models have been used to produce CFD-approximated solutions. Mokhtar et al. used cGANS to predict the wind flow on urban area patches (2). However, the authors highlighted the necessity of more extensive datasets and point out the unavailability of public urban flow databases(2).

The present work is intended to create a model capable to generate accurate two-dimensional (2D) CFD-like predictions for any urban geometry without needing to perform preliminary CFD studies on it. To do so, a vast dataset with more than  $20 \times 10^3$  samples has been generated to train a CNN that takes urban geometrical and meteorological parameters as input and produces the corresponding wind flow map for

the specified conditions.

## 2 Dataset Generation

Good Generalization capabilities are one of the main goals of the proposed model. To achieve this target, a careful selection of urban geometries has been carried out. The purpose is to obtain a rich dataset containing a wide range of flow conditions and geometrical features typical of urban areas.

### 2.1 Geometry selection criteria

Thirty actual European urban areas of 1 km<sup>2</sup> have been selected. Statistical analysis has been carried out to select the geometries, ensuring their representativity of the typical urban configurations that can be found in European cities. Specifically, k-means clustering has been performed from geometrical and land use parameters. The geometries showing less uniform distributions and the less cross-correlated ones (which ensures geometrical variety) have been selected.

### 2.2 High-fidelity wind flow simulations

Then, high-fidelity simulations were performed on the 30 selected geometries with three different wind directions, resulting in 90 micrometeorological conditions. Wall-Modeled Large Eddy Simulation has been performed to ensure accurate mean flow data. The Vreman model has been used as a subgrid model, while an equilibrium wall function with roughness has

