

Abstract Submitted
for the DFD17 Meeting of
The American Physical Society

**Improving urban wind
flow predictions through data assimilation**¹ JORGE SOUSA, Stanford Univ,
CATHERINE GORLE, Stanford University — Computational fluid dynamic is
fundamentally important to several aspects in the design of sustainable and resilient
urban environments. The prediction of the flow pattern for example can help to de-
termine pedestrian wind comfort, air quality, optimal building ventilation strategies,
and wind loading on buildings. However, the significant variability and uncertainty
in the boundary conditions poses a challenge when interpreting results as a basis for
design decisions. To improve our understanding of the uncertainties in the models
and develop better predictive tools, we started a pilot field measurement campaign
on Stanford University’s campus combined with a detailed numerical prediction of
the wind flow. The experimental data is being used to investigate the potential use
of data assimilation and inverse techniques to better characterize the uncertainty in
the results and improve the confidence in current wind flow predictions. We consider
the incoming wind direction and magnitude as unknown parameters and perform
a set of Reynolds-averaged Navier-Stokes simulations to build a polynomial chaos
expansion response surface at each sensor location. We subsequently use an inverse
ensemble Kalman filter to retrieve an estimate for the probabilistic density function
of the inflow parameters. Once these distributions are obtained, the forward analysis
is repeated to obtain predictions for the flow field in the entire urban canopy and
the results are compared with the experimental data.

¹We would like to acknowledge high-performance computing support from Yellow-
stone (ark:/85065/d7wd3xhc) provided by NCAR

Jorge Sousa
Stanford Univ

Date submitted: 01 Aug 2017

Electronic form version 1.4