

### 1. Motivation

Easy to use, efficient, and open source frameworks for assimilating data from multiple sensors currently do not exist.

PyDDA is an expandable framework that integrates data from weather radars and forecasting models using SciPy's optimization package to create meteorological fields.

### 2. Variational framework

Data from varying sensors can be assimilated using the 3D variational framework [1,2]. A cost function related to the difference between the retrieved field and that from a sensor  $j$  is minimized:

$$J(\mathbf{v}) = c_1 J_1(\mathbf{v}) + c_2 J_2(\mathbf{v}) + \dots + c_n J_n(\mathbf{v})$$

Since an initial state of the meteorological field must be prescribed, this is commonly taken from model/rawinsonde data, or set to zero. In addition, a  $J_n$  can also represent a physical constraint, such as conservation of mass. The  $c_n$  represent the relative importance of each sensor or constraint to  $J$ .

### 3.

PyDDA (Pythonic Dual Doppler Analysis) is a Python package that implements the 3D variational framework built on the Python ARM Radar Toolkit (Py-ART) [3]. Currently features:

- do multiple doppler analysis of an arbitrary number of radars using mass continuity, vorticity equations
- use Weather Research and Forecasting (WRF) data as initial state
- use rawinsonde data as initial state
- visualize wind fields w/wind barbs + streamlines

The limited memory Broyden–Fletcher–Goldfarb–Shanno - Box optimization technique from SciPy used for optimization:

```
V = fmin_l_bfgs_b(J, winds, args=(...),
                maxiter=10, pgtol=1e-3, bounds=bounds,
                fprime=grad_J, disp=1, iprint=-1)
```

PyDDA available for public use and contribution here:

<https://github.com/rcjackson/PyDDA>

### Example: Assimilating data from 2 radars + WRF model and visualize

#### Load two gridded radar files:

```
import pydda
import pyart
```

```
grid1 = pyart.io.read_grid(grid_name1)
grid2 = pyart.io.read_grid(grid_name2)
```

#### Then, prescribe the initial state from a WRF run:

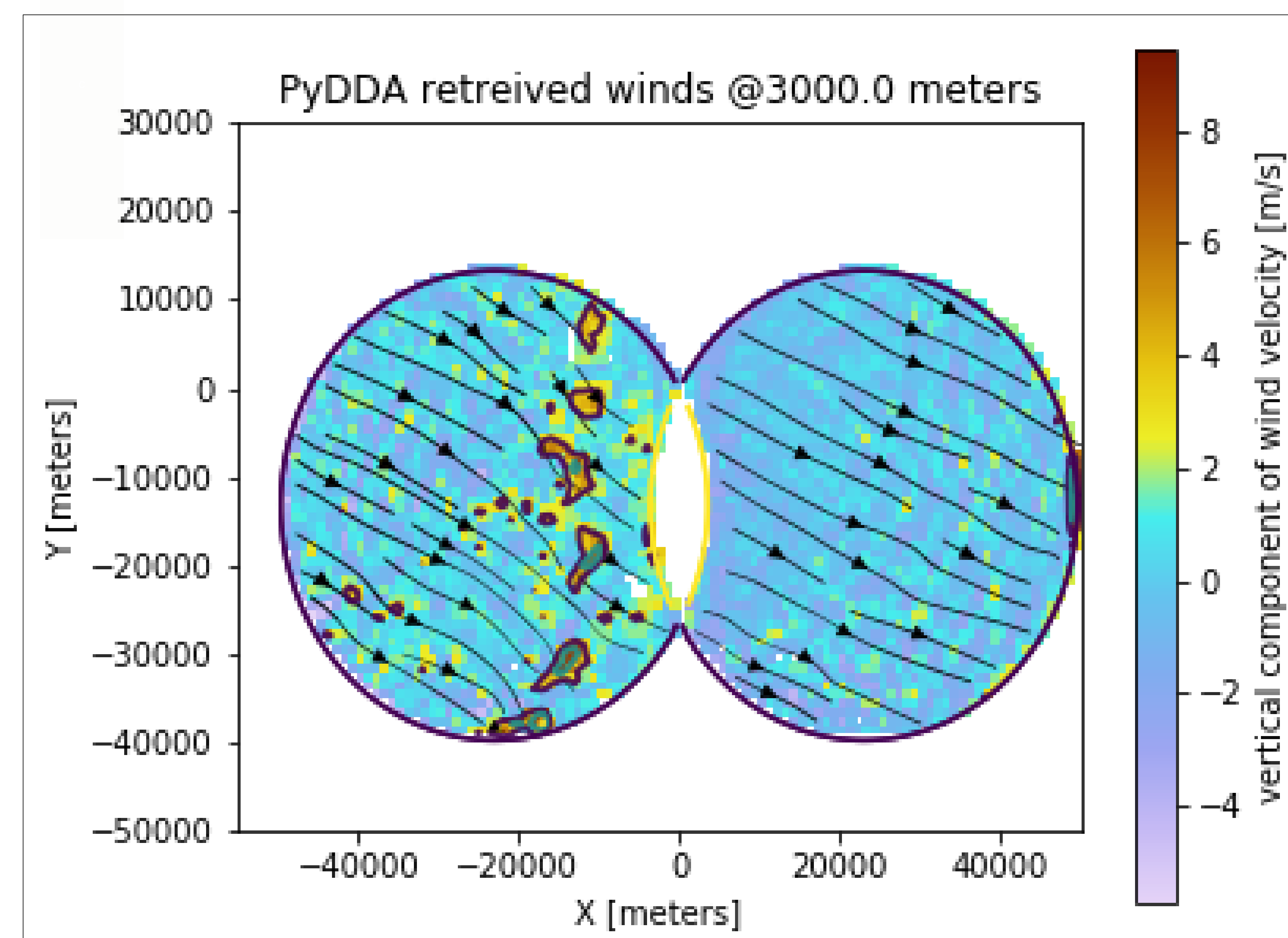
```
u1, v1, w1 = pydda.initialization.make_background_from_wrf(
    grid1, wrf_run_path, your_time, radar_loc, vel_field)
```

#### Let PyDDA retrieve the wind field:

```
Grids = pydda.retrieval.get_dd_wind_field(
    [grid1, grid2], u1, v1, w1, Co=100.0, Cm=1500.0)
```

#### Then, plot the result!

```
fig, ax = plt.subplots(1)
pydda.vis.plot_horiz_xsection_streamlines(
    Grids, ax, 'w', level=6, w_vel_contours=[3, 6, 9])
```



### References

- [1] Potvin, C.K., A. Shapiro, and M. Xue, 2012: Impact of a Vertical Vorticity Constraint in Variational Dual-Doppler Wind Analysis: Tests with Real and Simulated Supercell Data. J. Atmos. Oceanic Technol., 29, 32–49, <https://doi.org/10.1175/JTECH-D-11-00019.1>
- [2] Shapiro, A., C.K. Potvin, and J. Gao, 2009: Use of a Vertical Vorticity Equation in Variational Dual-Doppler Wind Analysis. J. Atmos. Oceanic Technol., 26, 2089–2106, <https://doi.org/10.1175/2009JTECHA1256.1>
- [3] Helmus, J.J. and Collis, S.M., 2016: The Python ARM Radar Toolkit (Py-ART), a Library for Working with Weather Radar Data in the Python Programming Language. Journal of Open Research Software. 4(1), p.e25. DOI: <http://doi.org/10.5334/jors.119>

We are looking for more collaborators to add more capabilities! If interested contact Bobby Jackson at [rjackson@anl.gov](mailto:rjackson@anl.gov).