

**We thank the reviewer for their time and their responses to our manuscript. Our replies are inserted below into the reviewer's comments.**

This manuscript provides the methodology and analysis behind a newly released publicly available offshore wind dataset called NOST (NOAAOffshoreWindProfiles-USA). The data set is six-hourly at 0.25-degree resolution and spans from 1987 to 2022. The authors implement a machine learning method (random forest regression) to extrapolate wind profiles from satellite-derived surface winds (NOAA's National Centers for Environmental Information's Blended Seawinds version 2.0 (NBSv2.0) product). They train their extrapolation model using lidar data and validate it at the same lidar sites as well as at two other locations whose data was not used for training. They compare their wind profiles with reference wind profiles (log-law, power law) against observations and find that their model outperforms reference wind profiles (as expected). They quantify their errors and also analyze subsets of the data to focus on low-level jets and high-shear conditions. Lastly, they use a triple collocation method to compare the NOSP product with ERA5 and the NOW23 wind dataset and find the lowest estimated errors for NOSP regardless of coastal region. The paper is very well-written, and I believe their dataset would be highly valued by wind farm developers. Overall, this is a great paper that can be improved with a few minor revisions as suggested below.

### **Specific Comments**

1. I would suggest removing or streamlining the first three sentences in the abstract. Considering that this is a wind energy-focused journal, the statements are already widely understood by the journal's target audience.

**The first three sentences were removed.**

2. I do not think that both the log-law and power law need to be used as references. I would recommend just choosing one as they are going to be similar and with similar limitations. **While we agree that only using one model for comparison may suffice, we believe that comparing against both is more thorough and readers may wonder how our model compares to the law we drop. As such we would prefer to leave both in if the reviewer finds that acceptable.**

Additionally, I could not find the surface roughness value used to generate the log-law profiles. Does the value vary spatially or temporally due to ocean conditions? The authors state that a lack of knowledge regarding  $u^*$  and  $z_0$  restrict them to only the neutral log-law, but you still need  $z_0$  for the neutral log-law so I am confused why  $u^*$  could not also be determined. I would also think that  $u^*$  is also an output of ERA5 similar to the SST.

**The surface roughness length used for the neutral log law is a constant  $z_0 = 0.0001$  (as in Optis et al., 2021). This has now been added into the paper (Line 180). As we used a constant  $z_0$ , it would not be possible to calculate  $u_*$  from  $z_0$  with any meaningful information as it would also be a constant.**

**$u_*$  is also available from ERA5 but is unavailable at the exact location of lidar stations used to train the model. As ERA5 has lower spatial (0.25° vs. in situ) and temporal resolutions**



**Thank you for finding this. The missing '%' is now added.**