

OFFSHORE ATMOSPHERIC STABILITY ESTIMATION FROM FLOATING LIDAR WIND PROFILES

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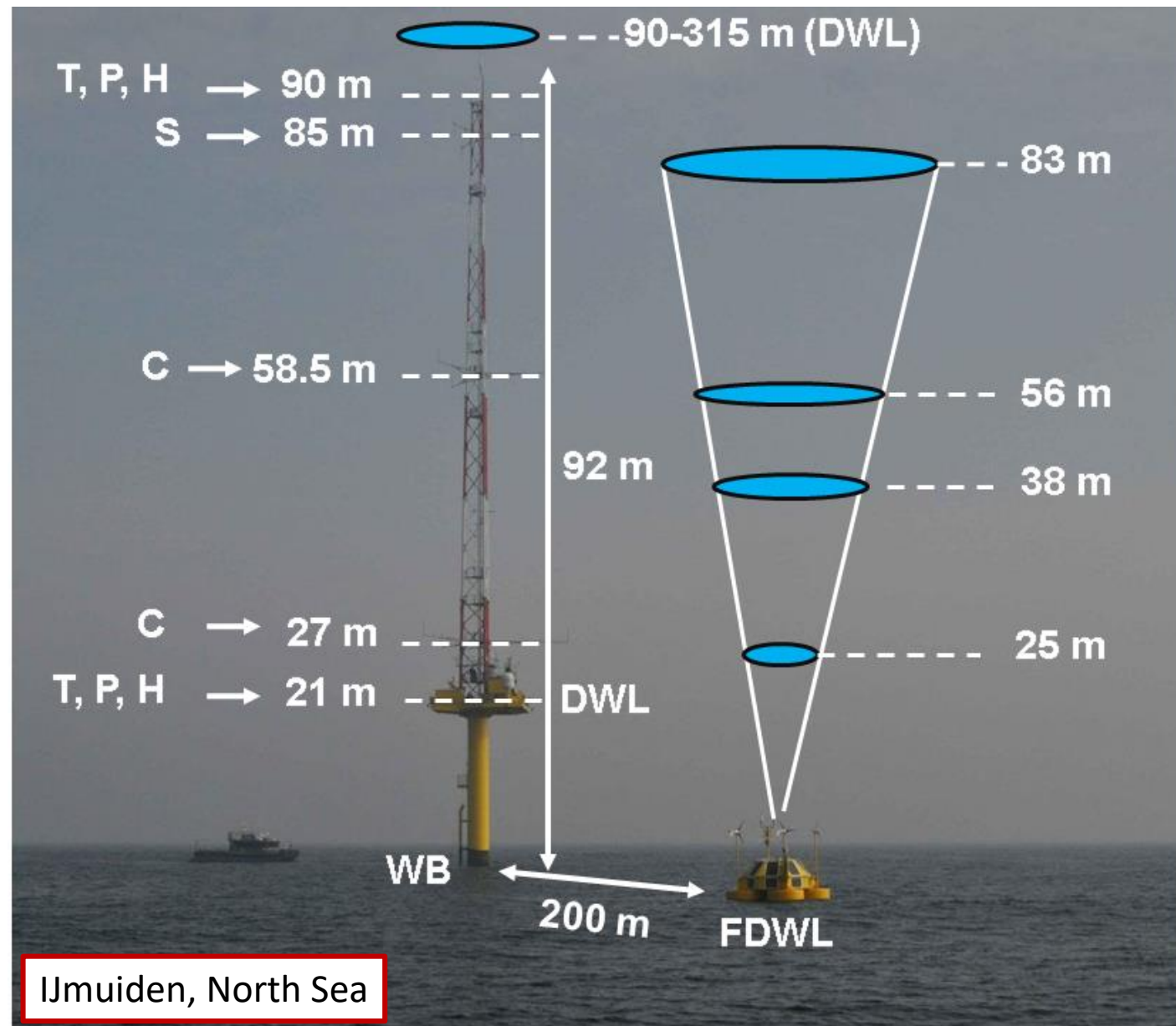
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OBJECTIVE

- Assess atmospheric stability from floating Doppler wind lidar (FDWL).
 - ✓ Power performance, wind shear, and wakes.
- Mast-measured reference parameters used to evaluate performance of the stability estimation algorithm.
- 82-day campaign → 10,822 10-min samples

SENSORS

FDWL → Zephir™ 300.
(T) Temperature sensor.
(P) Pressure sensor.
(H) Humidity.
(S) Sonic anemometer.
(C) Cup anemometer.
(WB) Wave and current buoy.



2D-algorithm estimated stability (FDWL)

FDWL data screening

- $HWS < 2$ or $HWS > 70$ [m/s]
- Spatial variation: confidence interval of 95%
- $N = 8263$ 10-min samples.

FDWL-measured
wind profileMOST wind-profile model;
Businguer-Dyer
functions: $U(z, u_*, L)$, $\vec{U}_{FDWL}(z)$
($z = 25, 38,$
56 and 85 m)Non-linear least squares
 $\arg \min_{L, u_*} |\vec{U}_{FDWL} - \vec{U}(z)|^2$

10-min

 $\hat{L}$ and $\hat{u}_*$

Araújo da Silva et al. (2022)

Reference stability (Mast)

Temperature observations from
sonic anemometers were not stored

Grachev and Fairall (1997)

 $T_{21} \rightarrow$ Temperature at 21 m
 $T_0 \rightarrow$ Temperature from
 wave buoy at 0 m
 $U_{27} \rightarrow$ HWS at 27 m
Bulk Richardson
numberDimensionless
stability, $\zeta(R_i)$ $z' = 15.5$ m $L_{Ri} = \frac{z'}{\zeta}$

Stability classification for the whole campaign

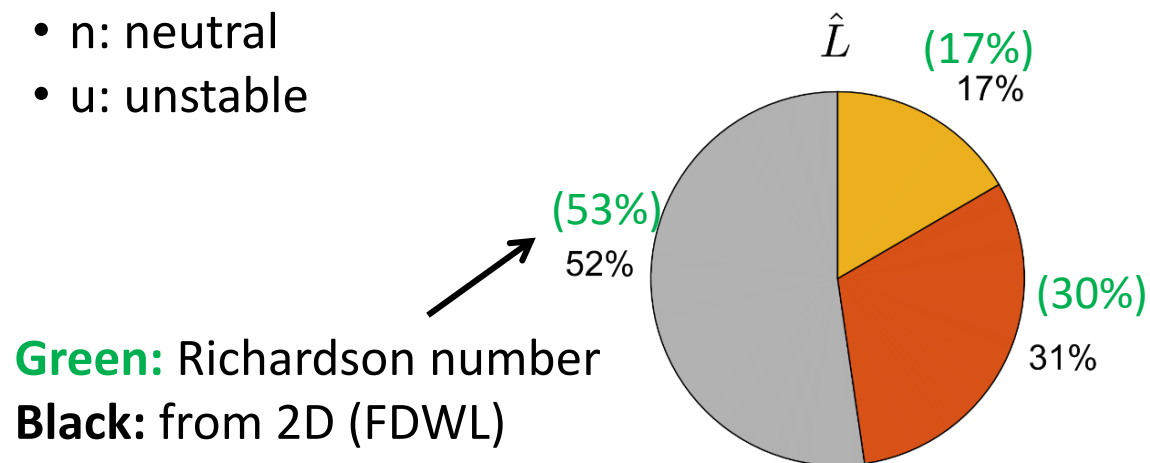
Table 1. Stability classes based on the Obukhov length, L .

Atmospheric Stability	Obukhov length range (m)
Stable	$10 < L < 500$
Neutral	$ L > 500$
Unstable	$-500 < L < -50$

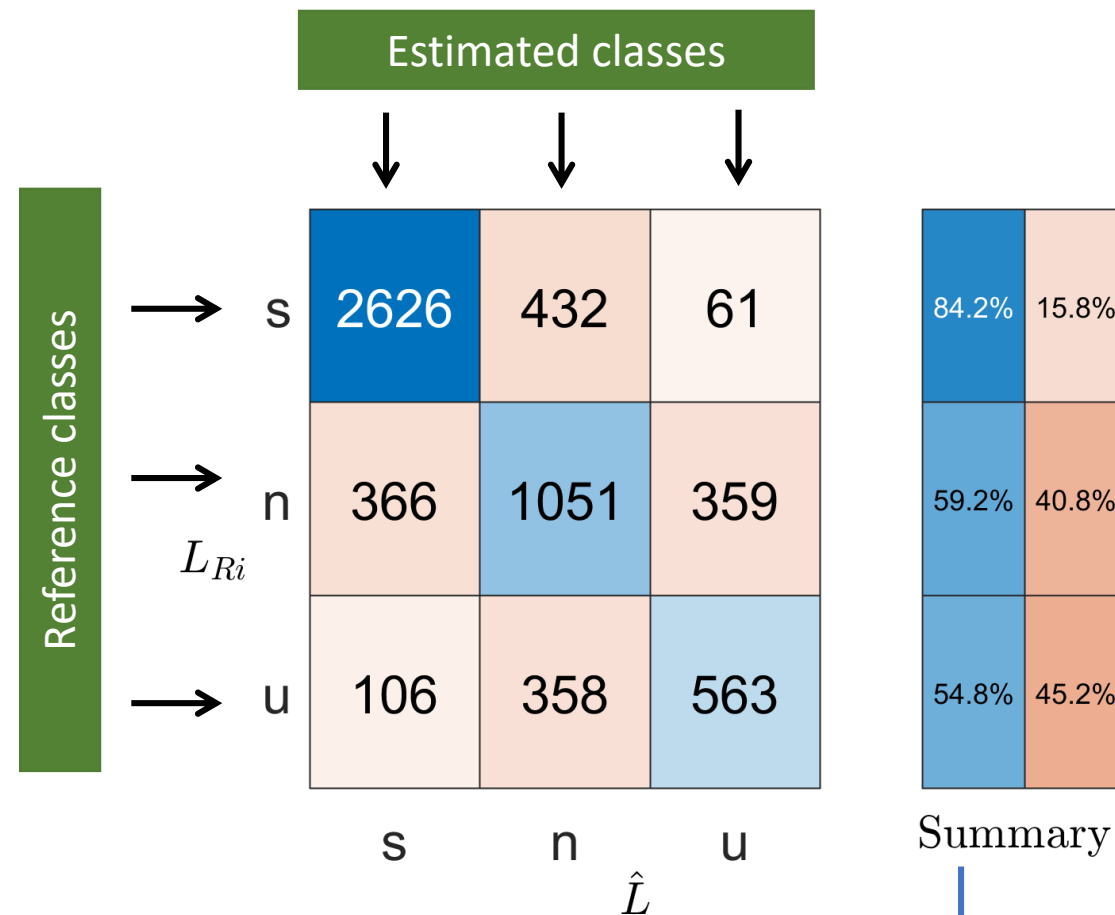
5,922 valid samples

Gryning et al. (2007)

- s: stable
- n: neutral
- u: unstable



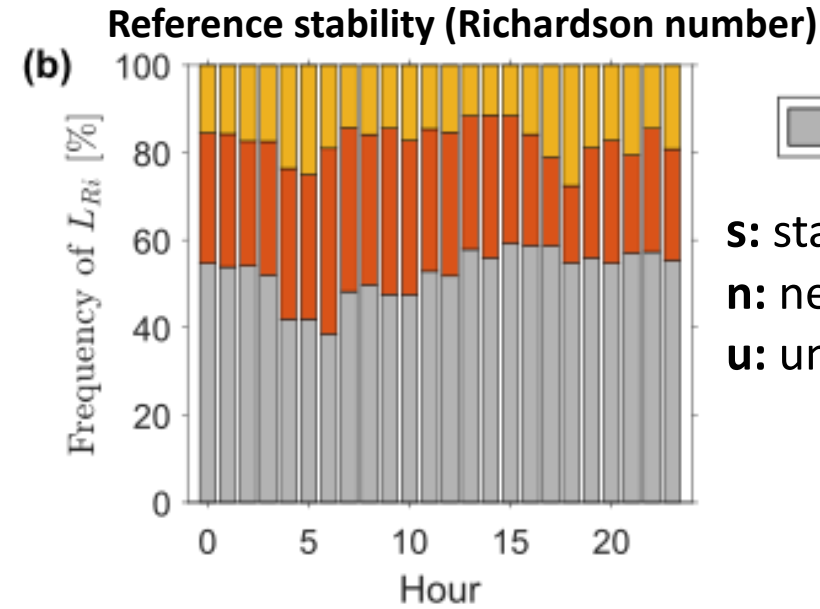
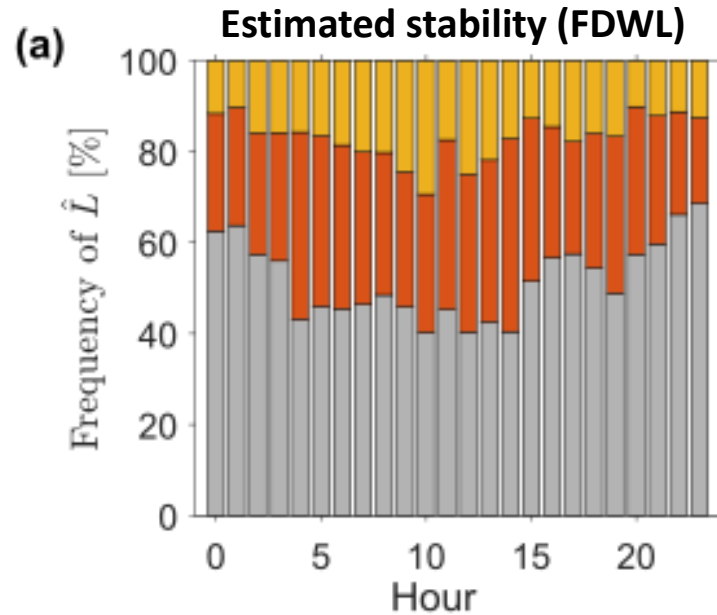
Confusion matrix



Summary

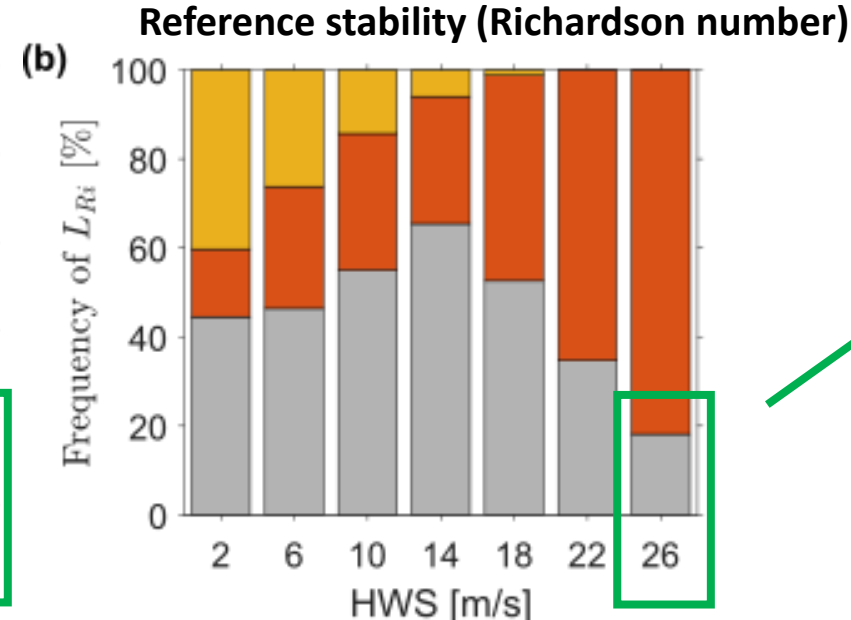
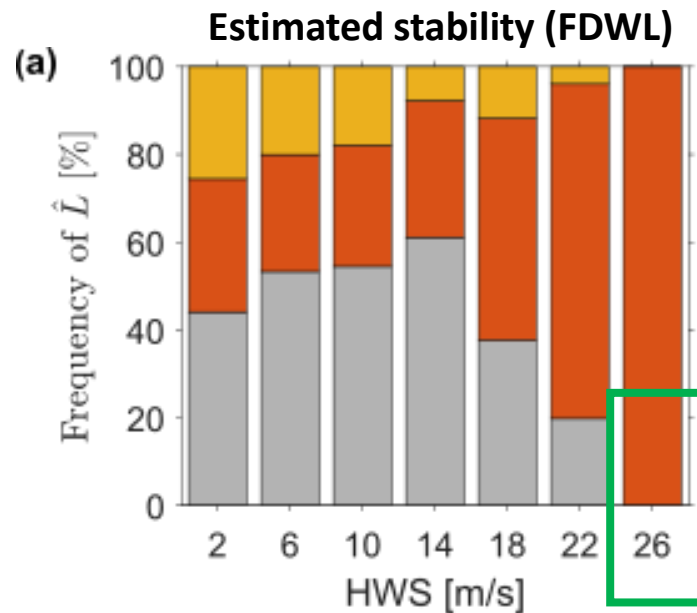
Mean hit rate: ~66%

Stability
classification
clustered by hour
of the day



s: stable $\rightarrow 10 < L < 500$
n: neutral $\rightarrow |L| > 500$
u: unstable $\rightarrow -500 < L < -50$

Stability
classification
clustered by
horizontal wind
speed (HWS)



Stable class is
underestimated
at 26-m/s cluster

- A 2-D algorithm for *atmospheric stability classification* was derived.
- The algorithm consists of fitting the MOST wind profile model to the FDWL-measured wind profile via non-linear LSQ optimization.
 - Relies on estimation the *friction velocity* and *Obukhov length** (*semi-quantitative*).
→ aiming to classify the atmospheric stability in *three classes*.
 - Confusion matrix → Mean HIT RATE = 66%.
- The 2-D algorithm enabled to assess:
 - The stability evolution during the time of day.
 - The horizontal-wind-speed dependence on the stability.

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PRESENTATION REFERENCES

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Thanks!

2D estimated stability (FDWL)

FDWL data screening

- $HWS < 2$ or $HWS > 100$ [m/s]
- Spatial variation: confidence interval of 95%
- $N = 8263$ 10-min samples.



FDWL-measured
wind profile

$$\alpha = 0.012$$

$$U(z) = \frac{u_*}{k} \left[\ln \left(\frac{gz}{\alpha u_*^2} \right) - \psi \left(\frac{z}{L} \right) \right]$$

$\vec{U}(z)$

$\vec{U}_{FDWL}(z)$
($z = 25, 38,$
56 and 85 m)

Non-linear least Squares
 $\arg \min_{L, u_*} |\vec{U}_{FDWL} - \vec{U}(z)|^2$

10-min

$\hat{L}$ and $\hat{u}_*$

Reference stability (Mast)

$T_{21} \rightarrow$ Temperature at 21 m
 $T_0 \rightarrow$ Temperature from
wave buoy at 0 m
 $U_{27} \rightarrow$ HWS at 27 m

$$Ri = \frac{g \Delta \bar{\theta}_v \Delta z}{\bar{\theta}_{v21} (\bar{U}_{27})^2}$$

$$\zeta = 10 Ri, \text{ if } Ri \leq 0$$

$$\zeta = \frac{10 Ri}{1 - 5 Ri}, \text{ if } Ri \geq 0$$

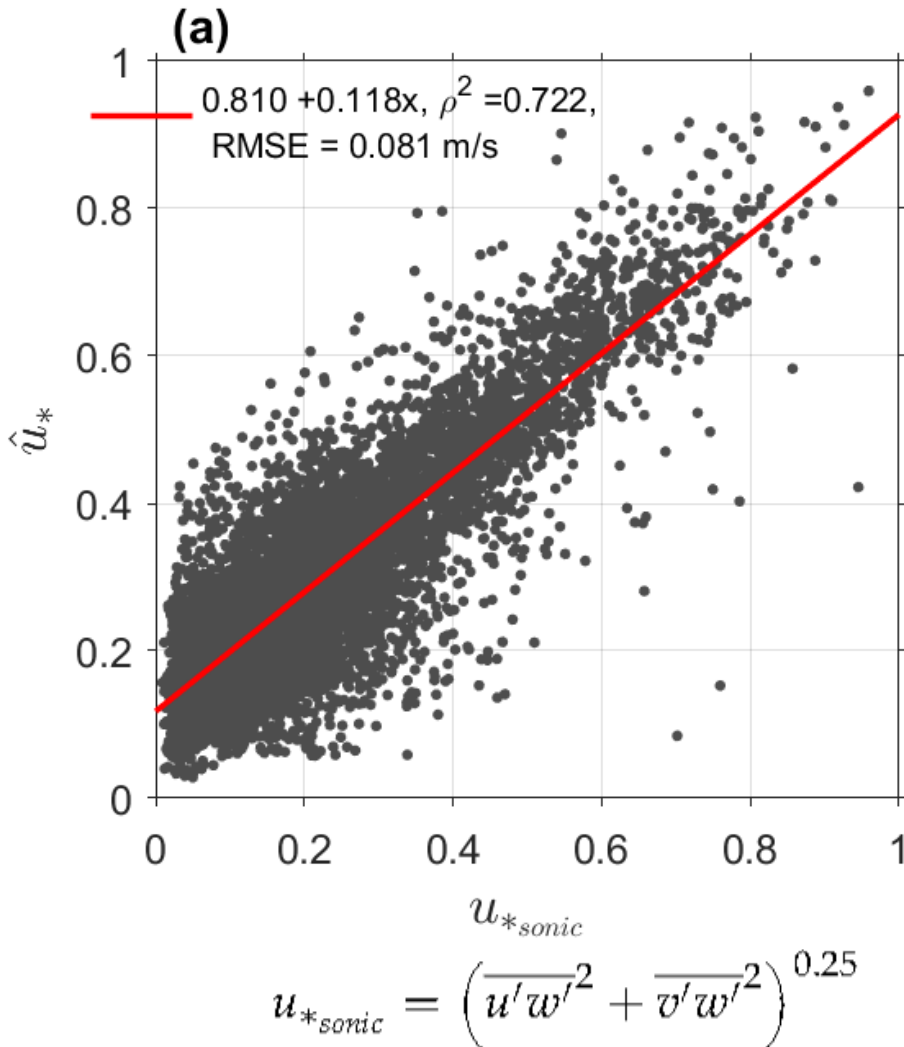
Grachev and Fairall (1997)

$$z' = 15.5 \text{ m}$$

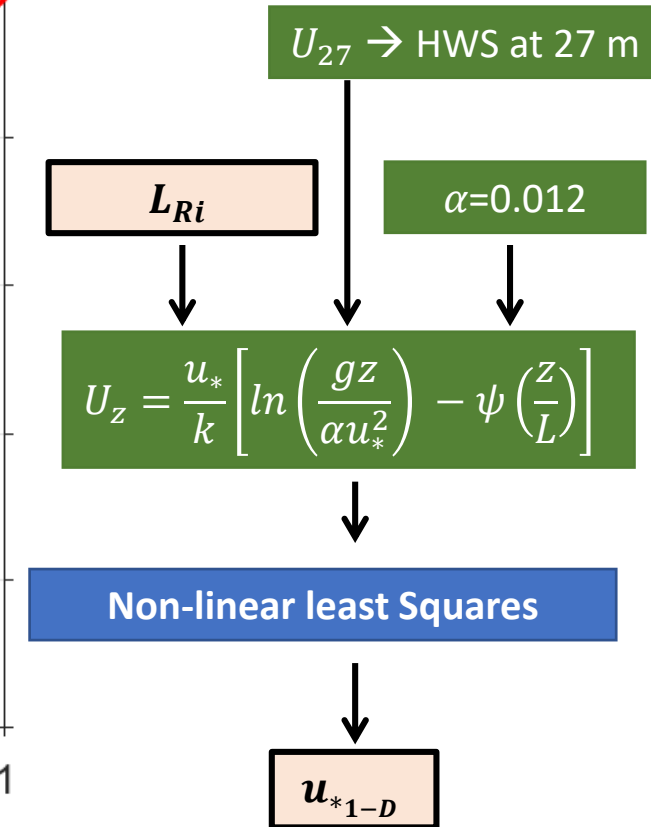
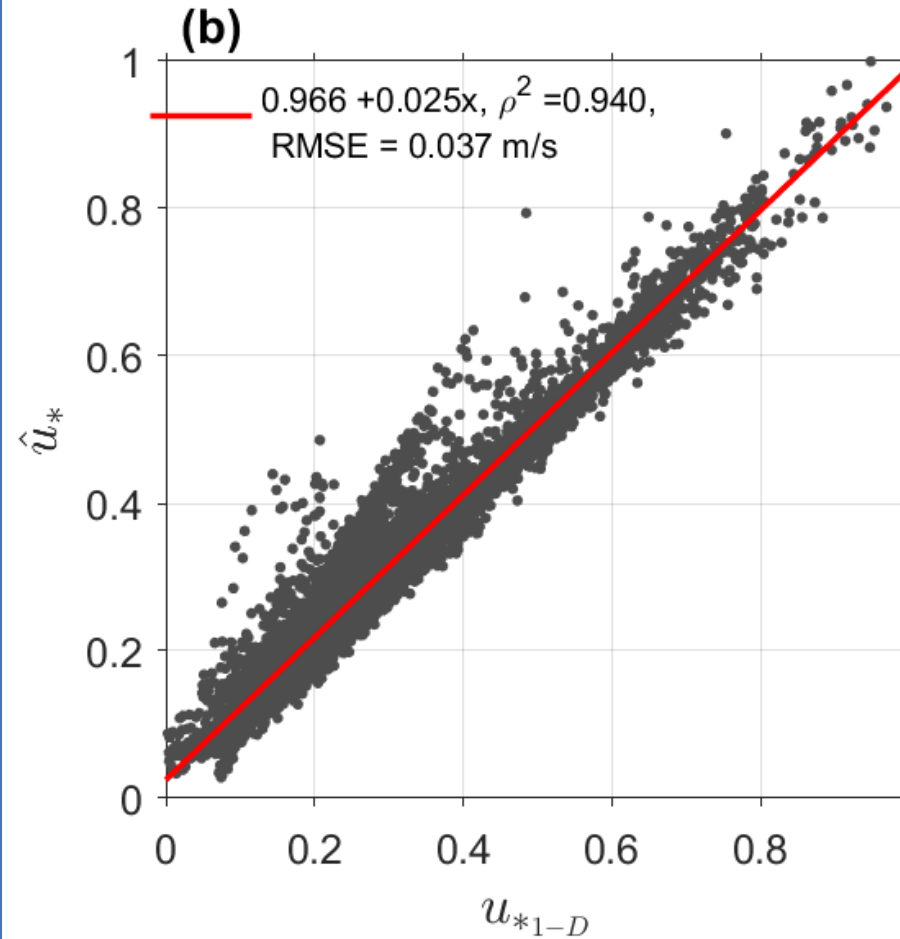
$$L_{Ri} = \frac{z'}{\zeta}$$

$T_{21} \rightarrow$ Temperature at 21 m
 $T_0 \rightarrow$ Temperature from
wave buoy at 0 m
 $U_{27} \rightarrow$ HWS at 27 m

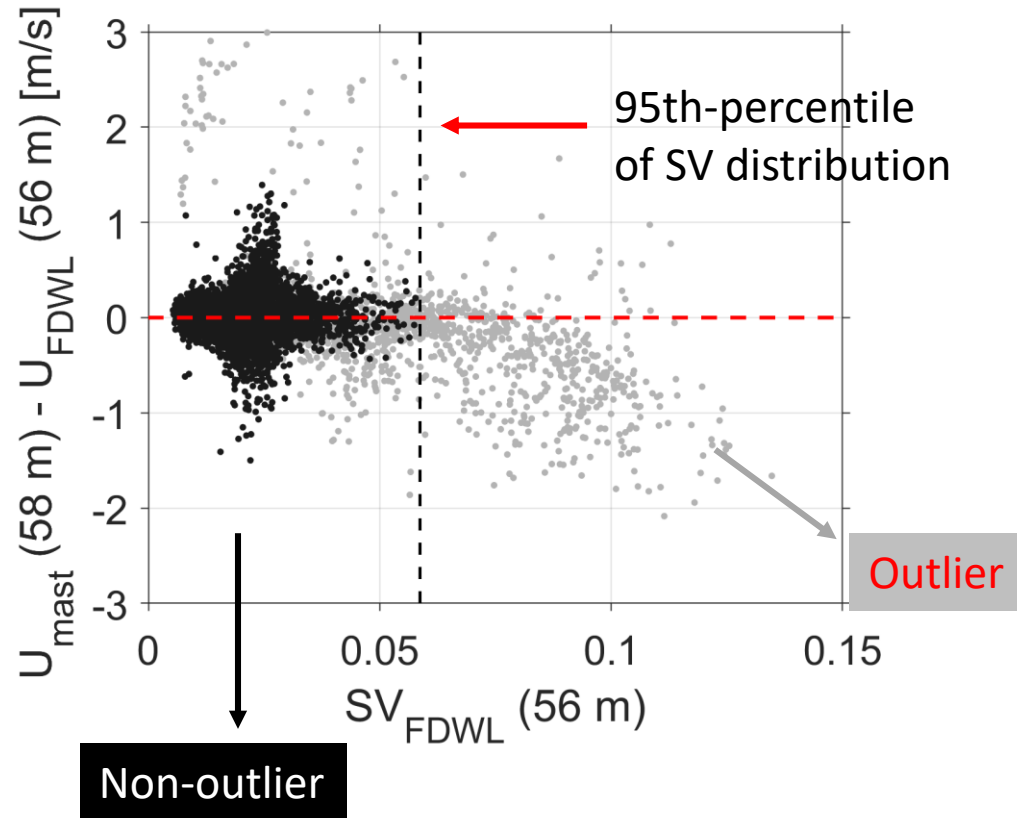
Reference: Sonic anemometers



Reference: u_* numerically solved using



Data screening

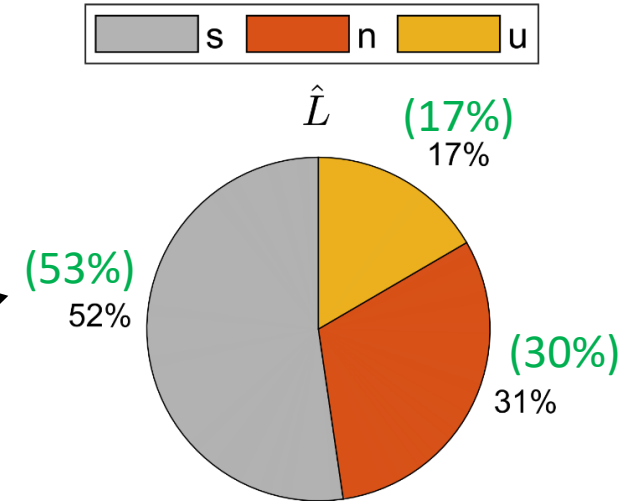


Valid samples → 95%
 Outlier samples → 5%
Spatial variation → lidar fitting quality

Stability classification for the whole campaign

- s: stable
- n: neutral
- u: unstable

Green: L from Richardson number
Black: L from FDWL

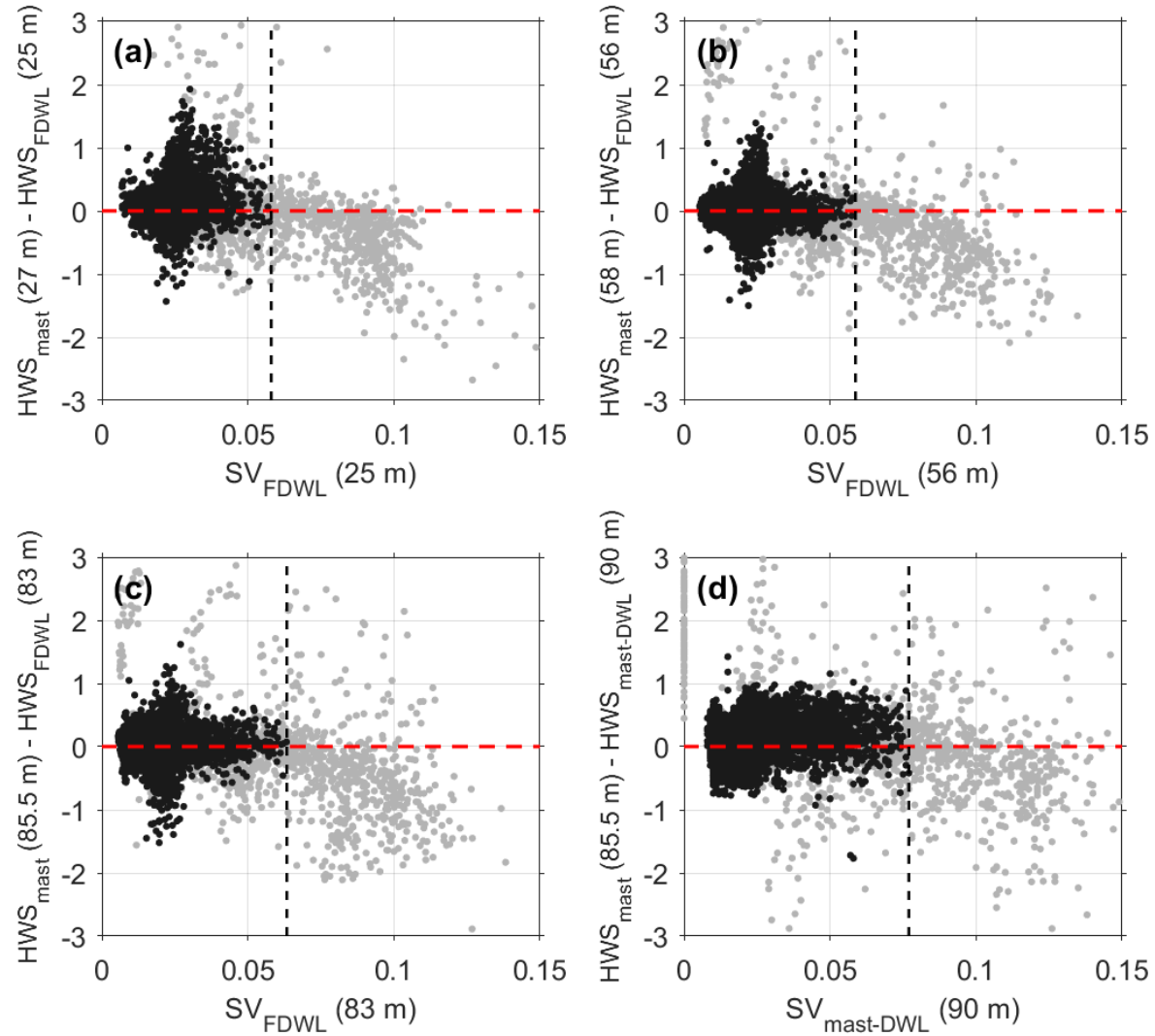


Confusion matrix

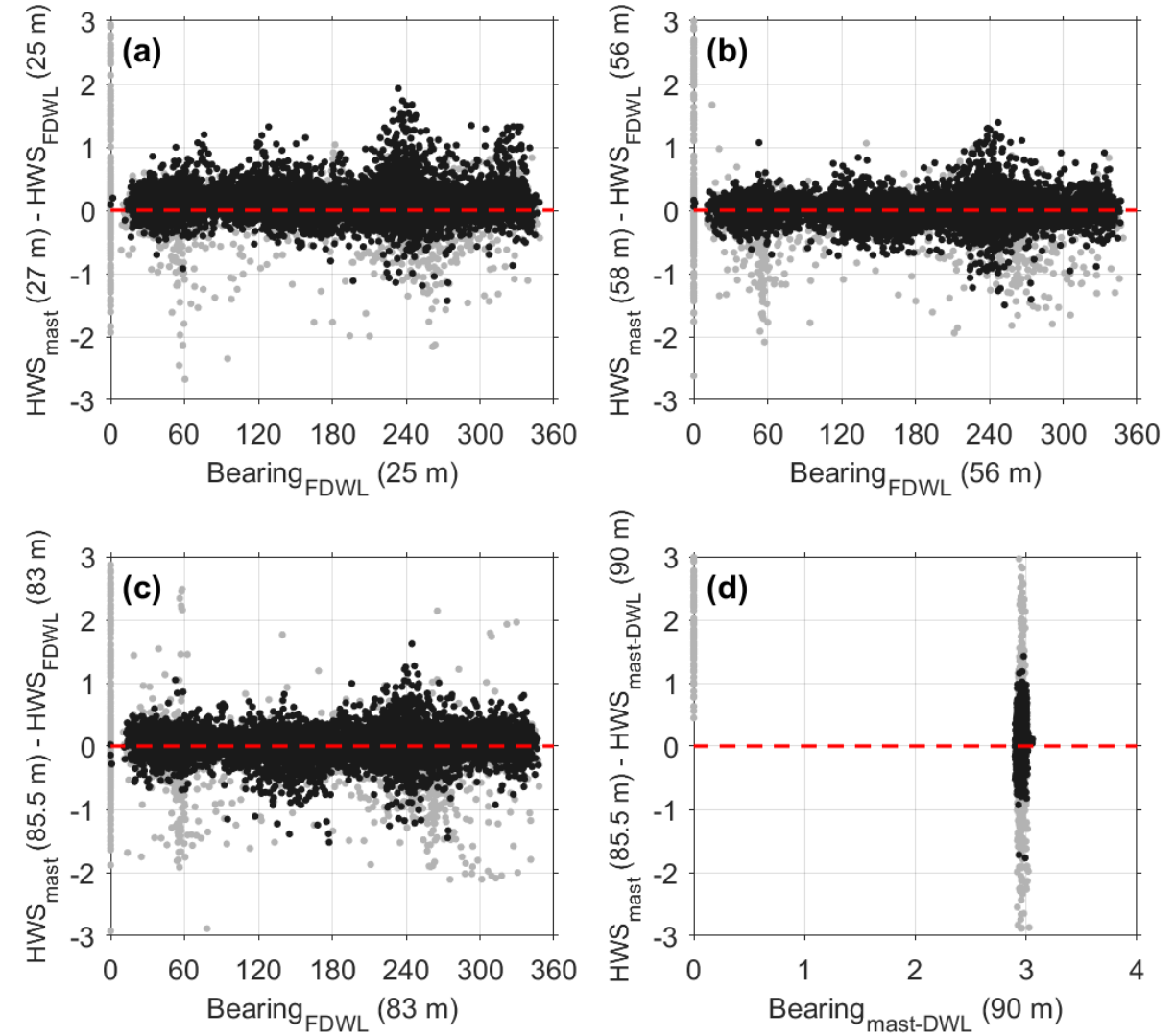
Mean hit rate: ~66%

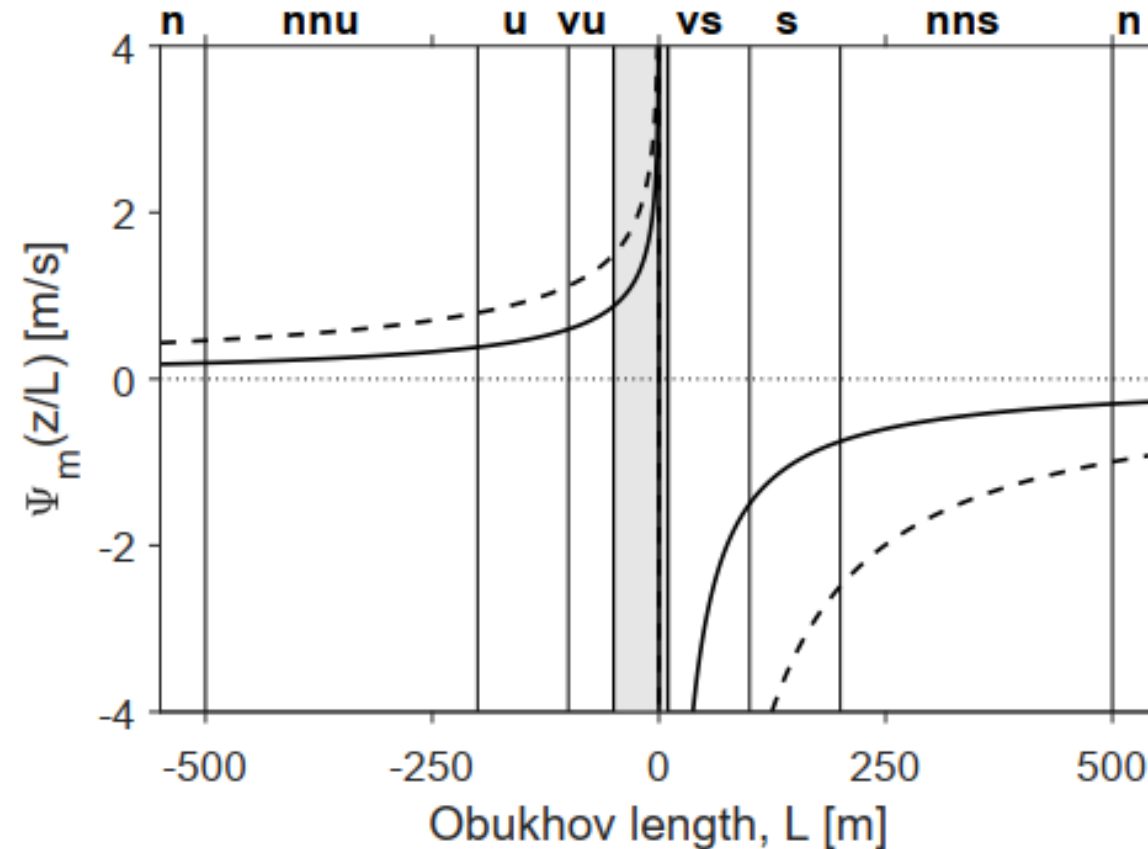
L_{Ri}	$\hat{L}$				
	s	n	u		
	s	n	u	Summary	
	s	n	u		
s	2626	432	61	84.2%	15.8%
n	366	1051	359	59.2%	40.8%
u	106	358	563	54.8%	45.2%

Data screening



Data screening





Araújo da Silva et al. (2022)

Figure 2. Stability-correction function $\Psi_m(\frac{z}{L})$ as function of Obukhov's length for $z = 27 \text{ m}$ (thick-solid line) and $z = 83 \text{ m}$ (dashed line). (Vertical grey-shaded) area delimits the interval from $L = -50 \text{ m}$ to $L = 10 \text{ m}$. (Vertical lines) delimit Gryning et al. [35] stability-classification thresholds as per Table 2. Labels: (vu) very unstable, (u), unstable, (nnu), near-neutral unstable, (n), neutral, (nns) near-neutral stable, (s) stable and (vs) is very stable.