

Measurements of Stratified Turbulence in Mountain Terrain. Bursting phenomena.

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Chris Hocut
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Motivation and Layout of the talk

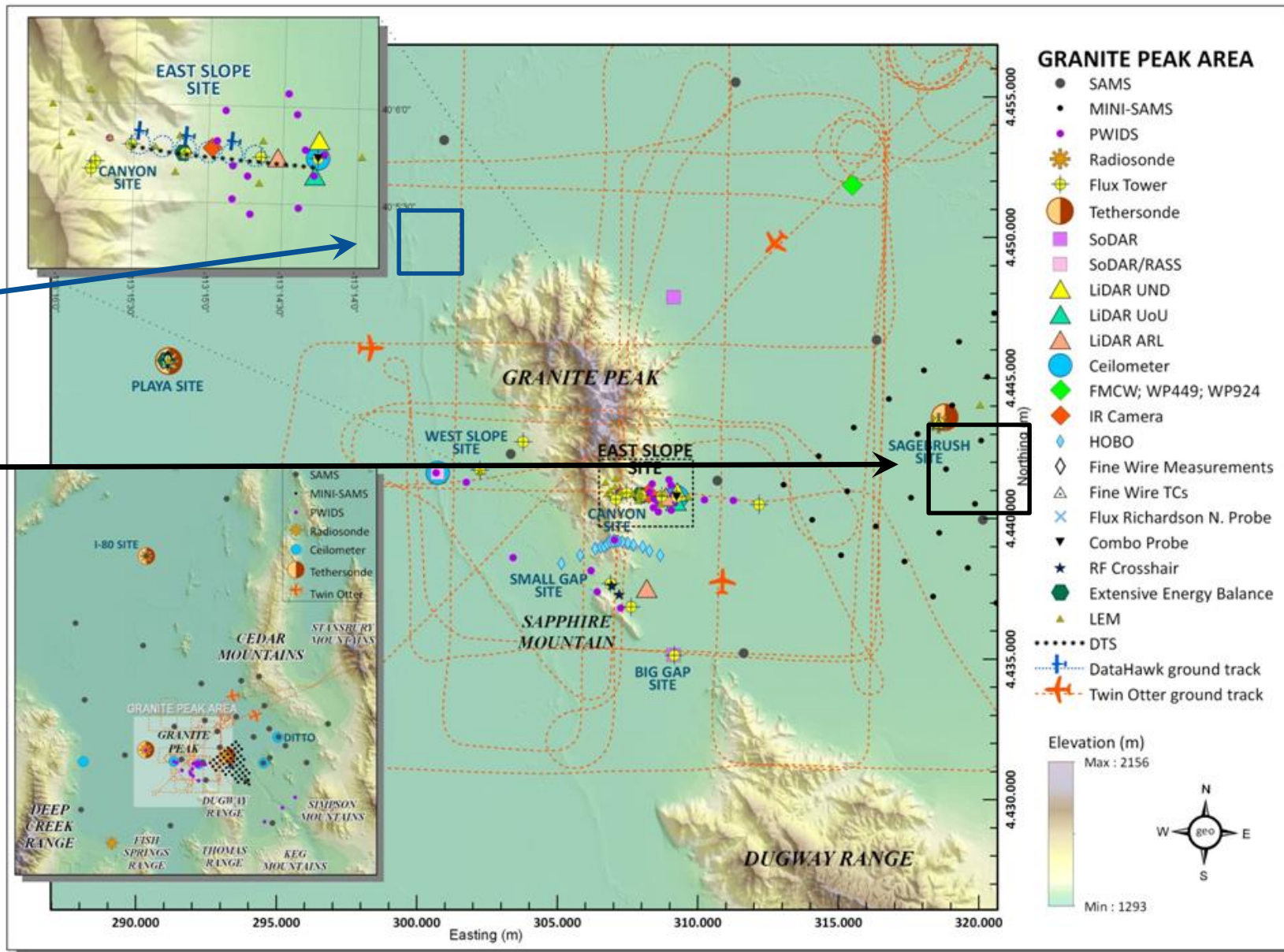
- Fine resolution measurements of atmospheric turbulence, which enable to determine dissipation, velocity derivatives etc. using a combo instrument (multi-sensor hot films embedded within a sonic).
- The use of in-situ calibration by utilizing a low resolution data from Sonic and NN procedure.
- Experimental results on stratified turbulence from fall campaign. Transition from stratified turbulence to Kolmogorov turbulence.
- Bursting phenomena. Internal waves and their breaking as path to bursting. Just thoughts.
- Conclusions

- 
- **DUGWAY FALL RESULTS**
 - **STRATIFIED TURBULENCE**
 - **BURSTING**

Combo Probe Placement

X-1

X-2



MATERHORN-X: ES2 tower.

Combos at 2 and 6 m

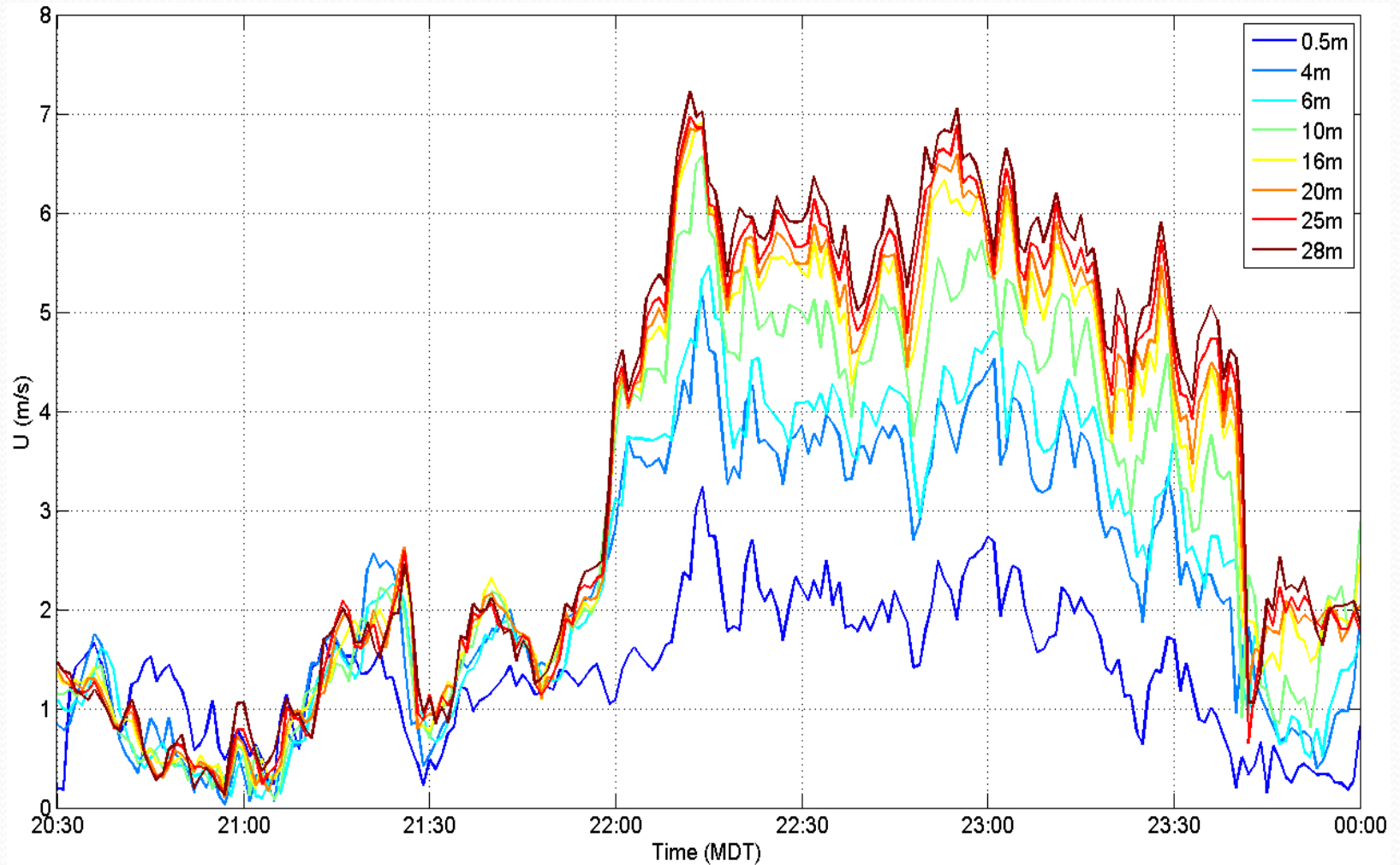


Combo probes located at 2m and 6m

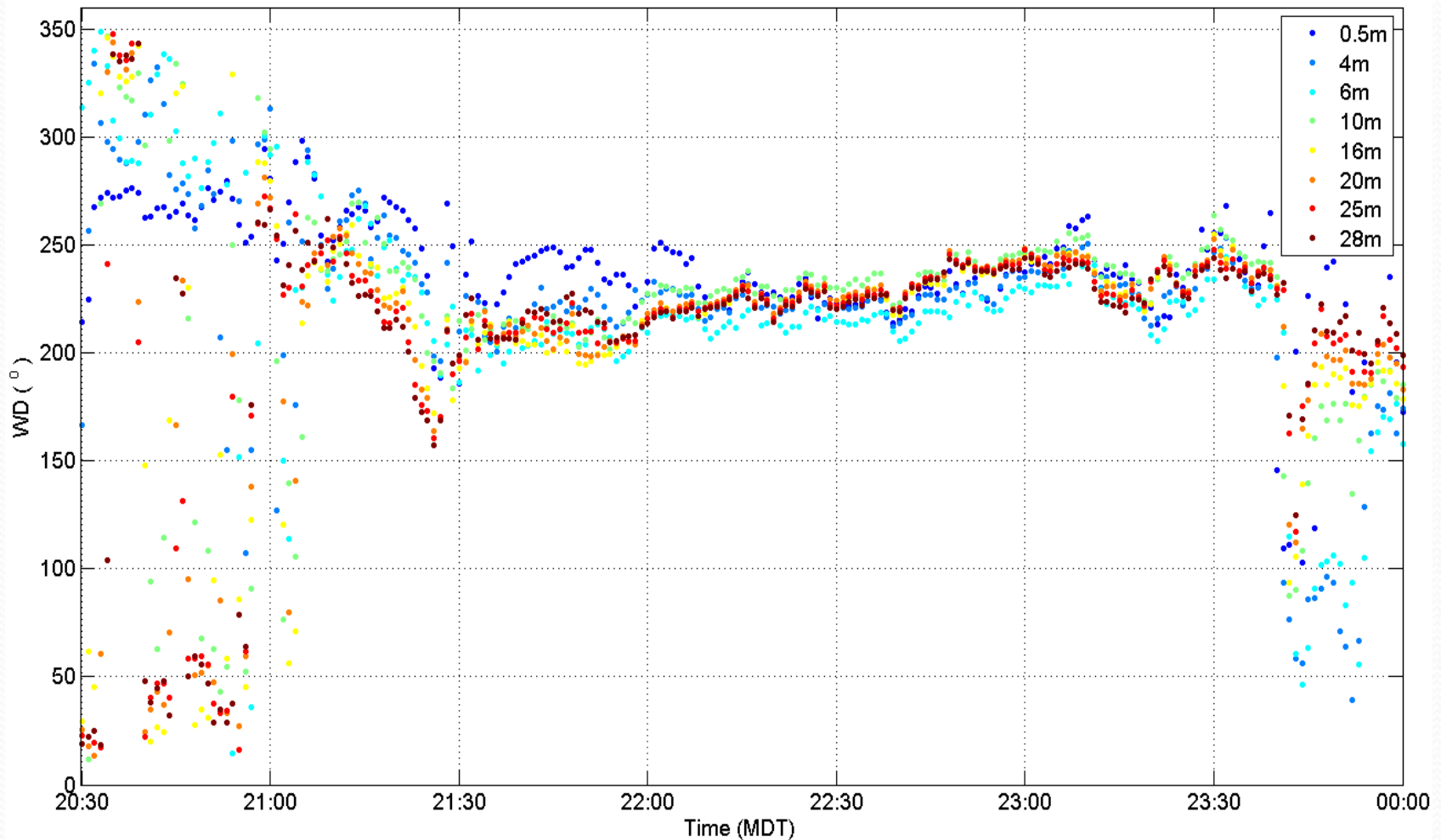


Combo probe electronics

Wind speed at the ES-2 tower (10/19/2012). At 22:00 MDT, wind speed rapidly increased as the flow developed.

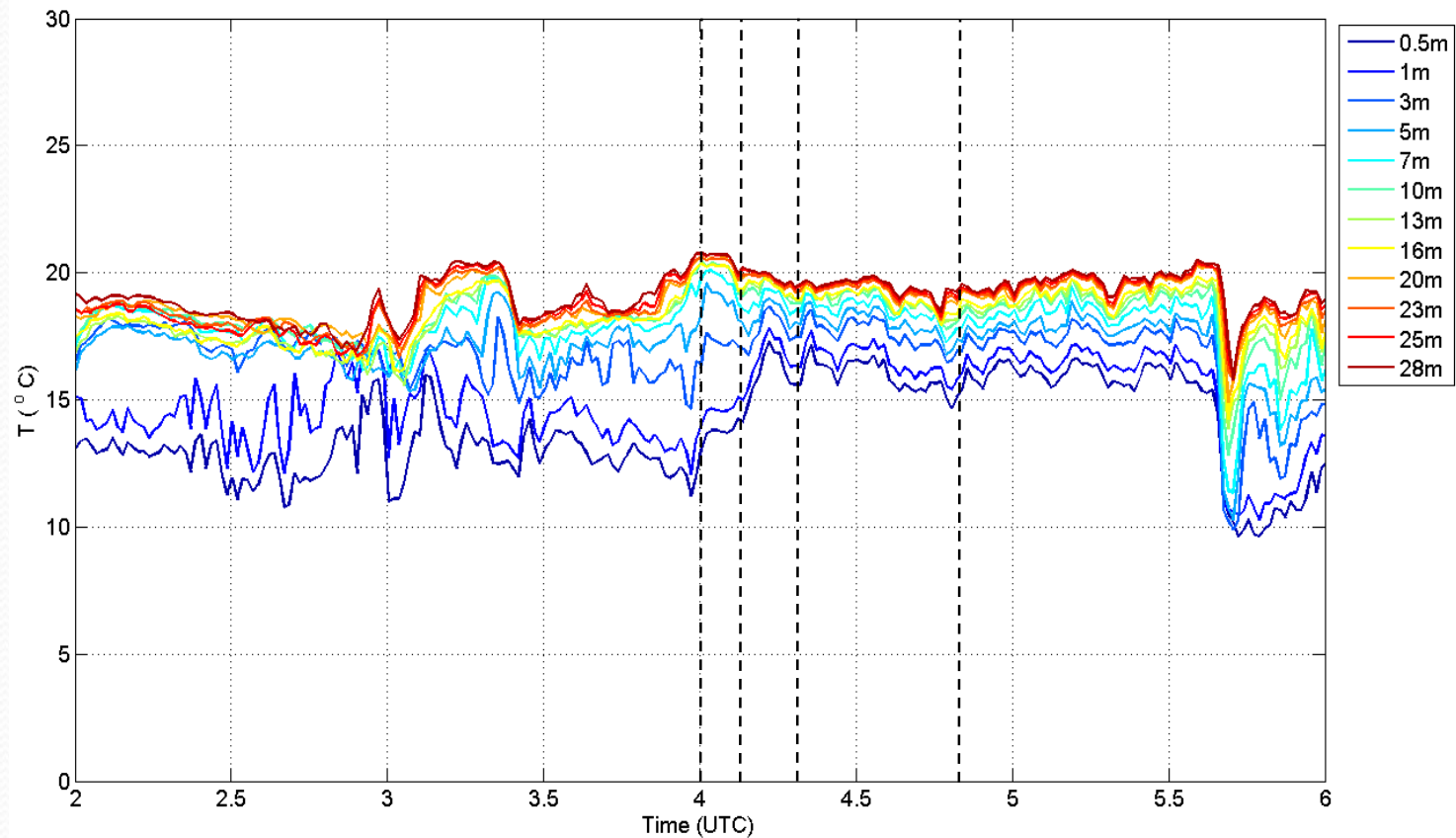


Wind direction at the ES-2 tower (10/19/2012). Prior to 21:00 MDT, the wind direction oscillated due to interacting katabatic and downvalley flow. After the flow developed the wind direction was nearly constant during the measurement period (22:00-23:30 MDT) throughout the height of the tower.

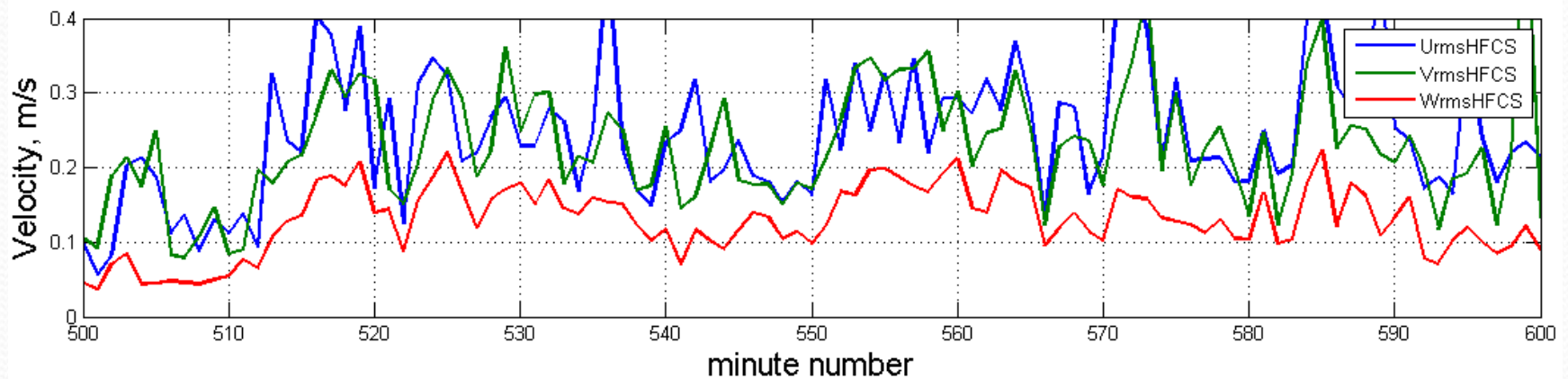
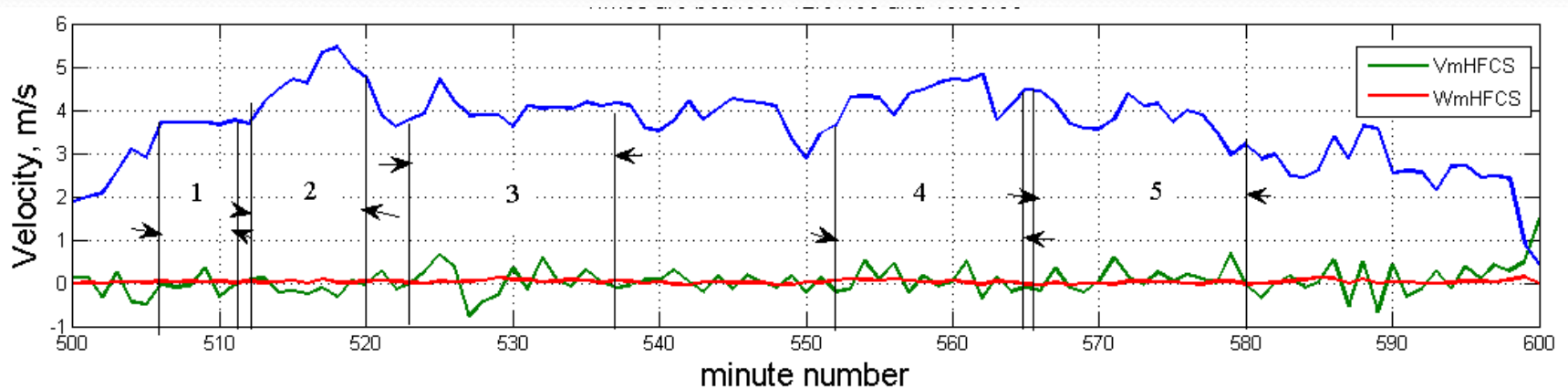


Temperature at ES2 tower (10/19/2012).

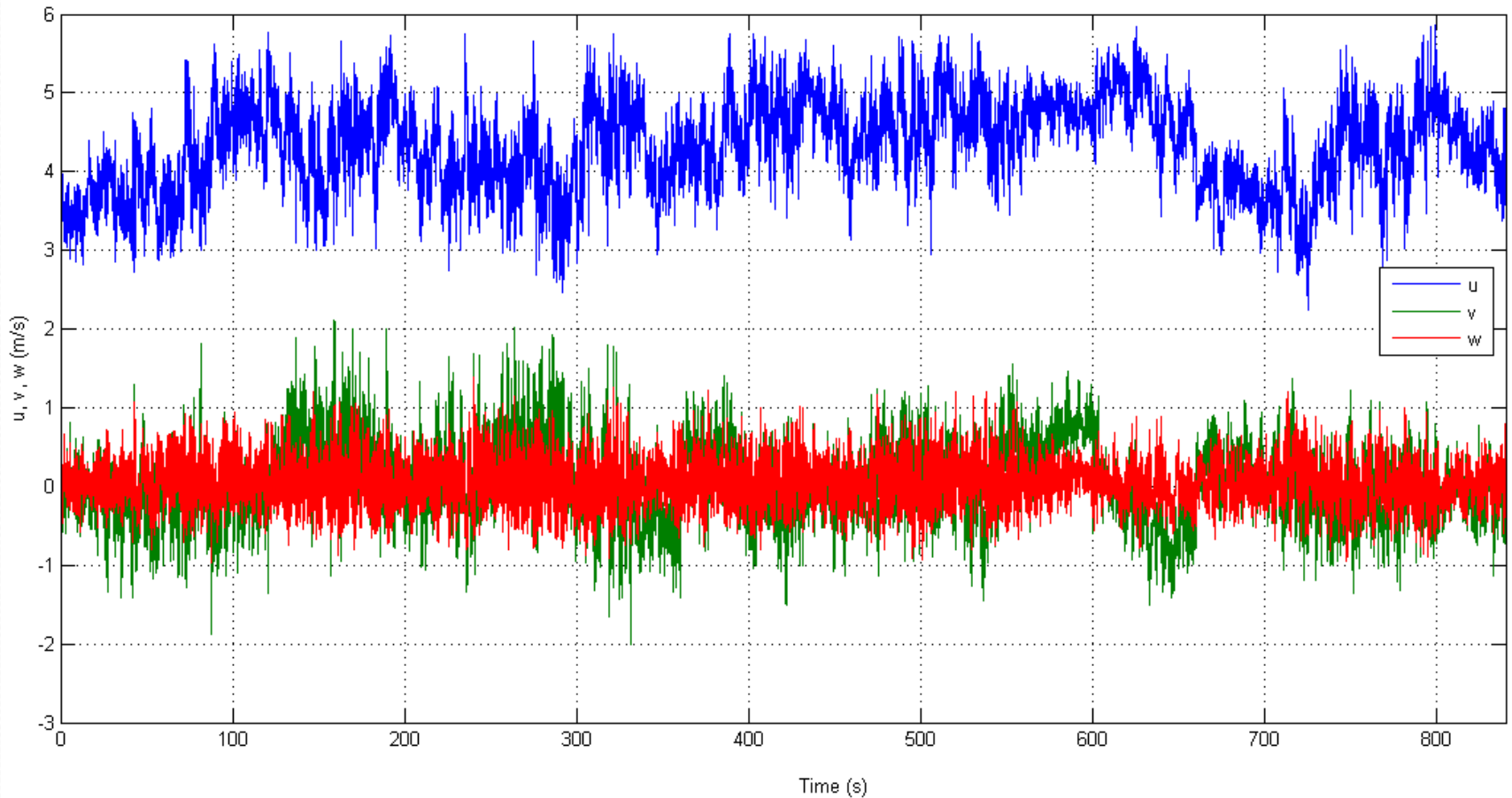
(Local Utah time: MDT = UTC-6).



Sonic time series for the nocturnal time period at October 19 (10:00-11:30PM)

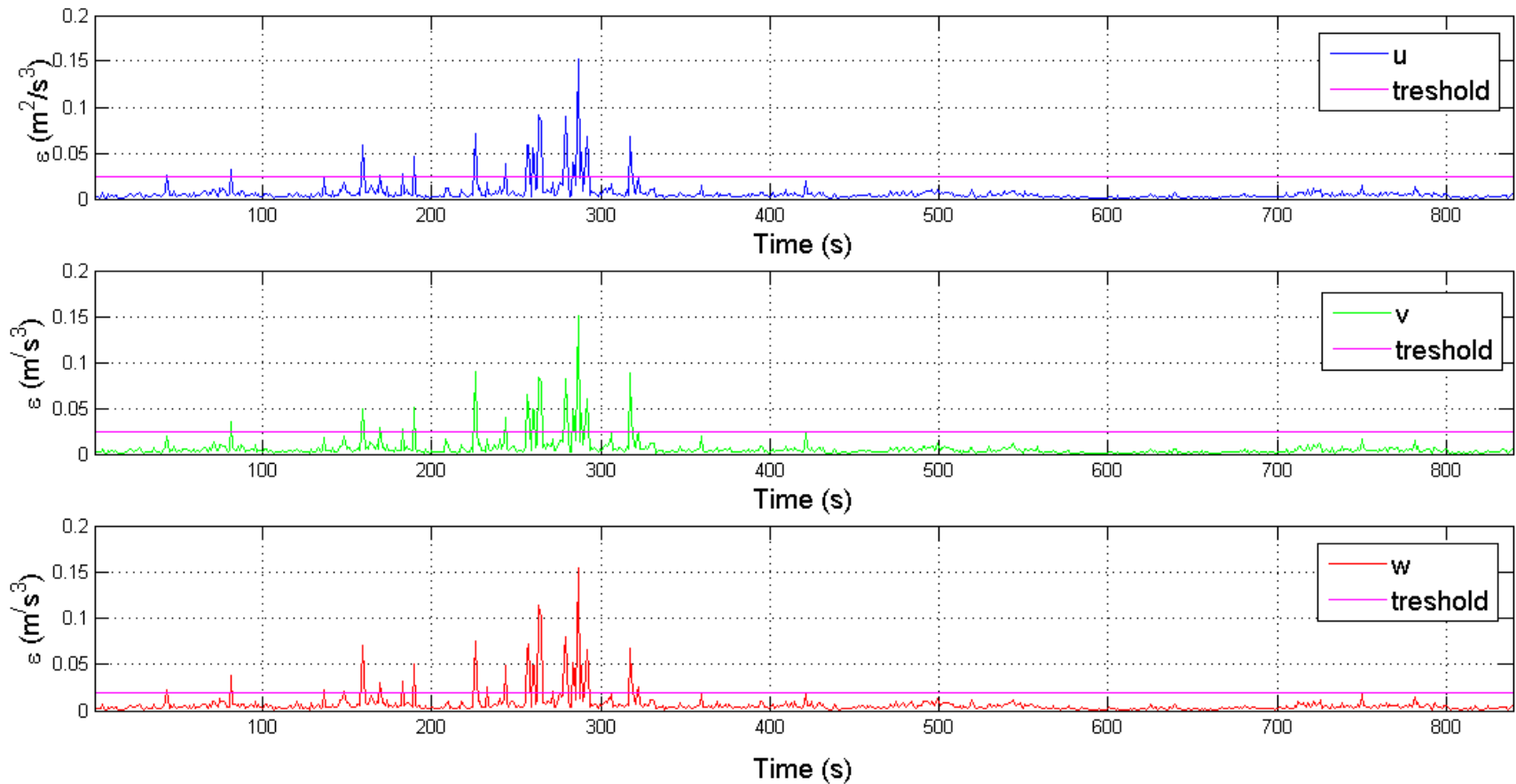


Fourth sub-interval (S-I). Reconstructed velocity field

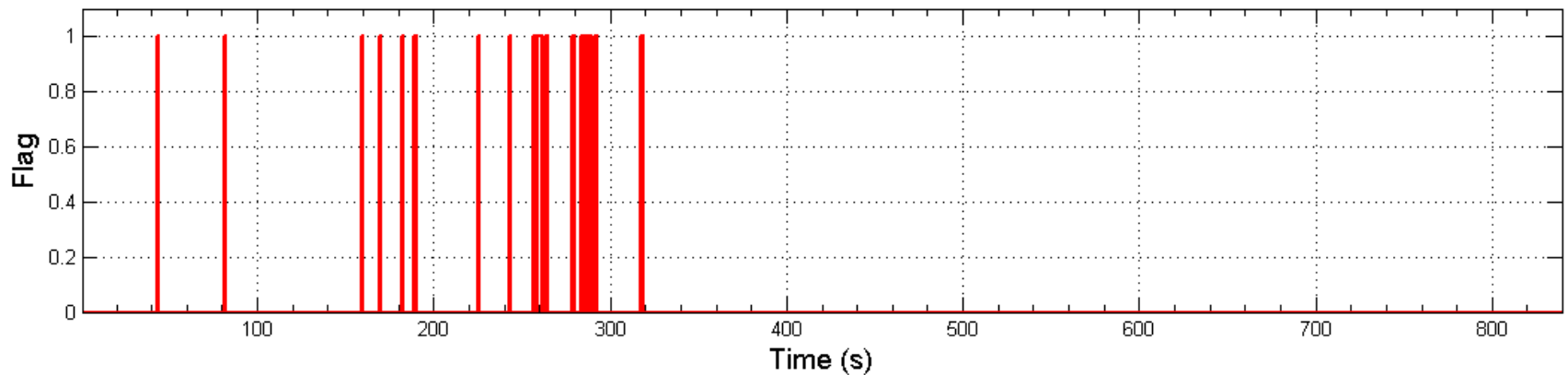
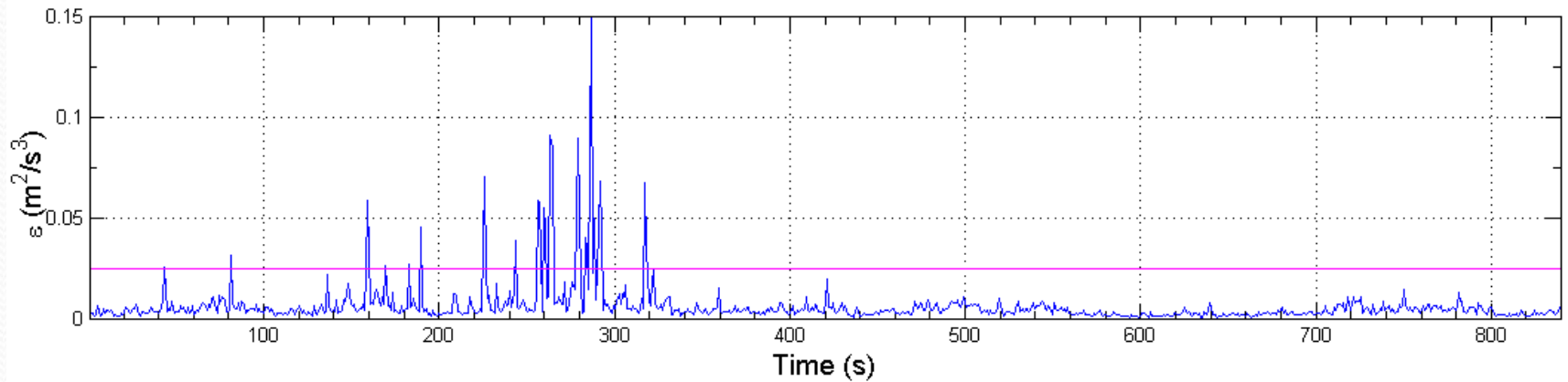


TKE dissipation in 4th S-I

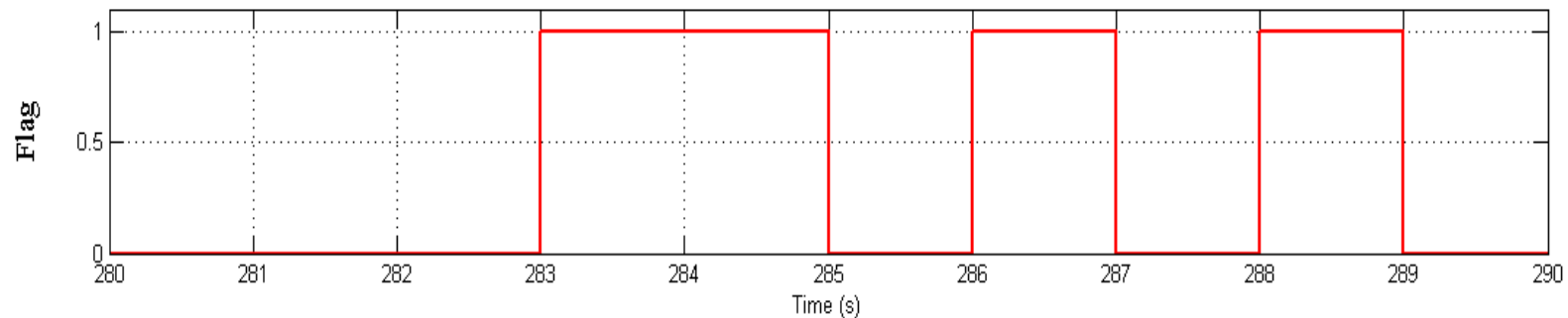
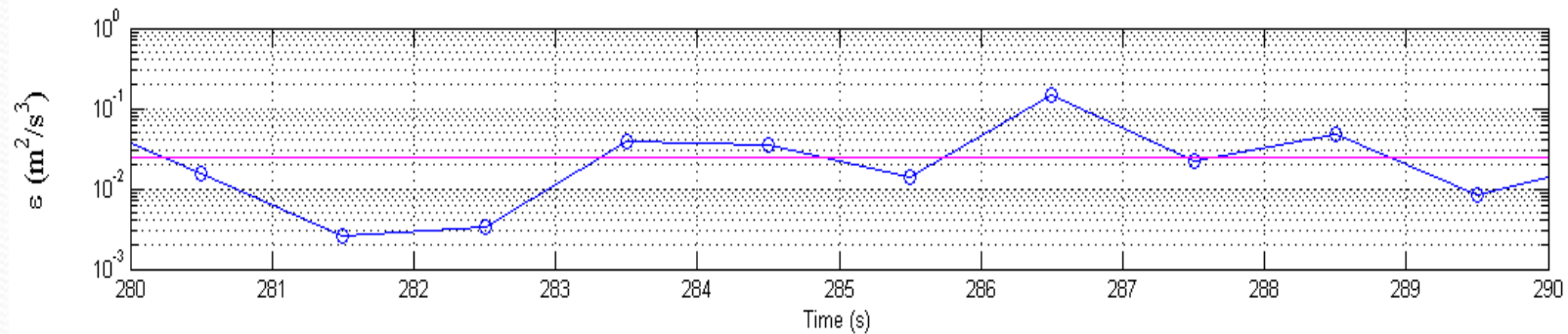
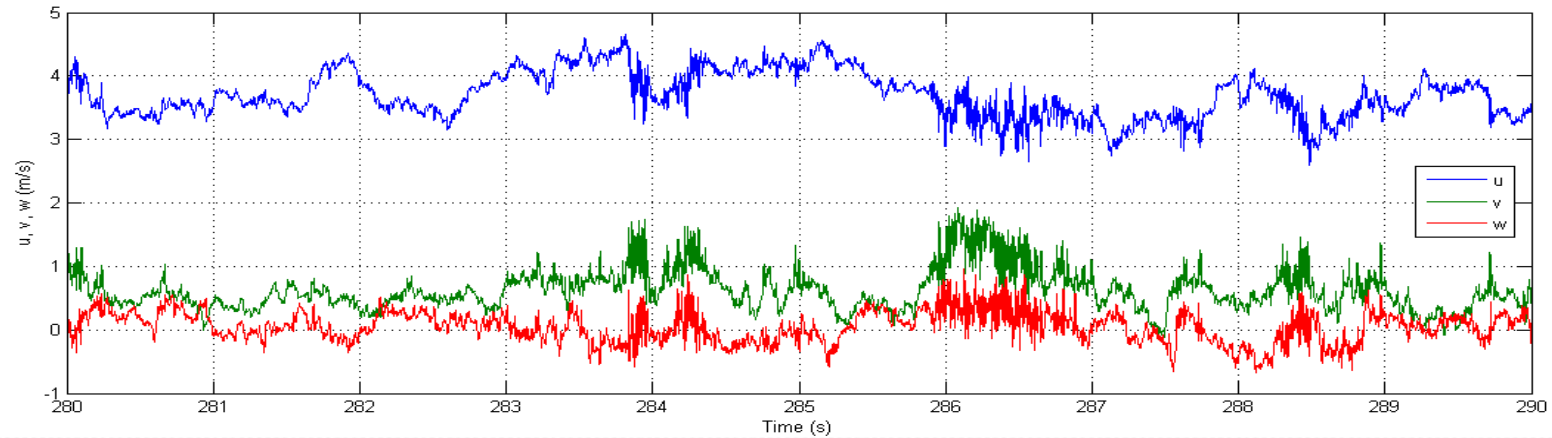
$$\epsilon = 15\nu \overline{\left(\frac{\partial u}{\partial x}\right)^2} ; \partial x = -U\partial t$$



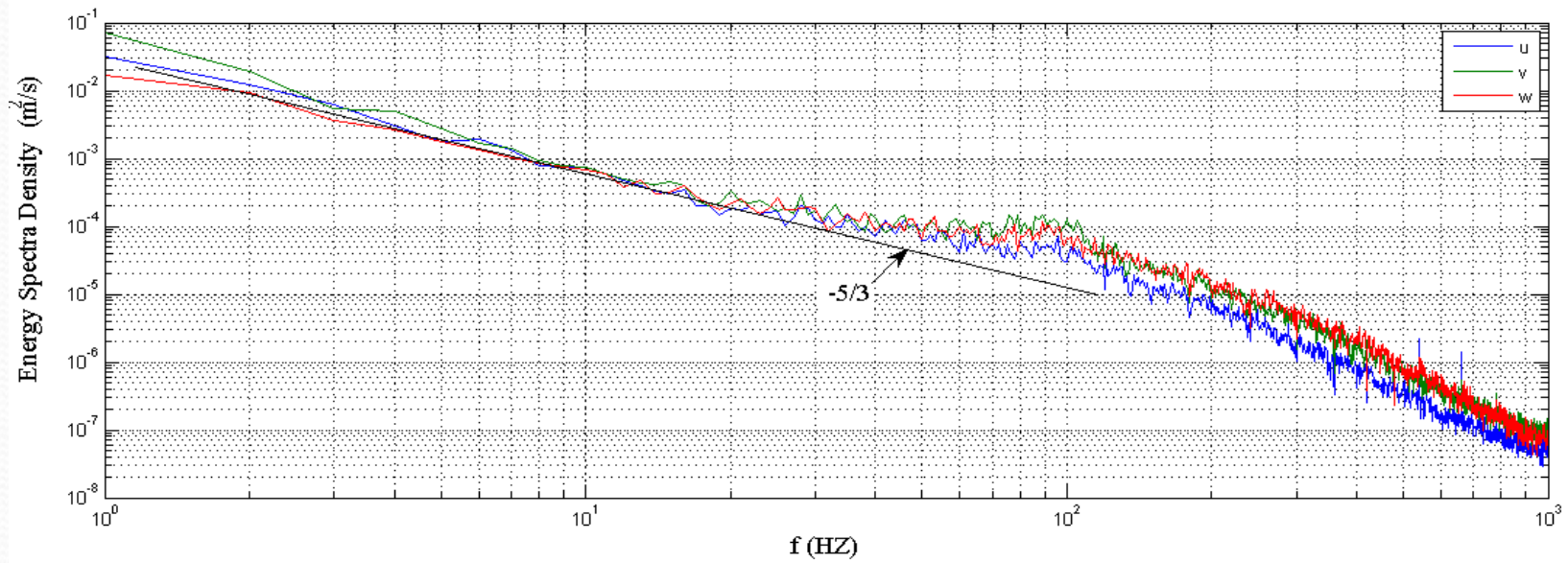
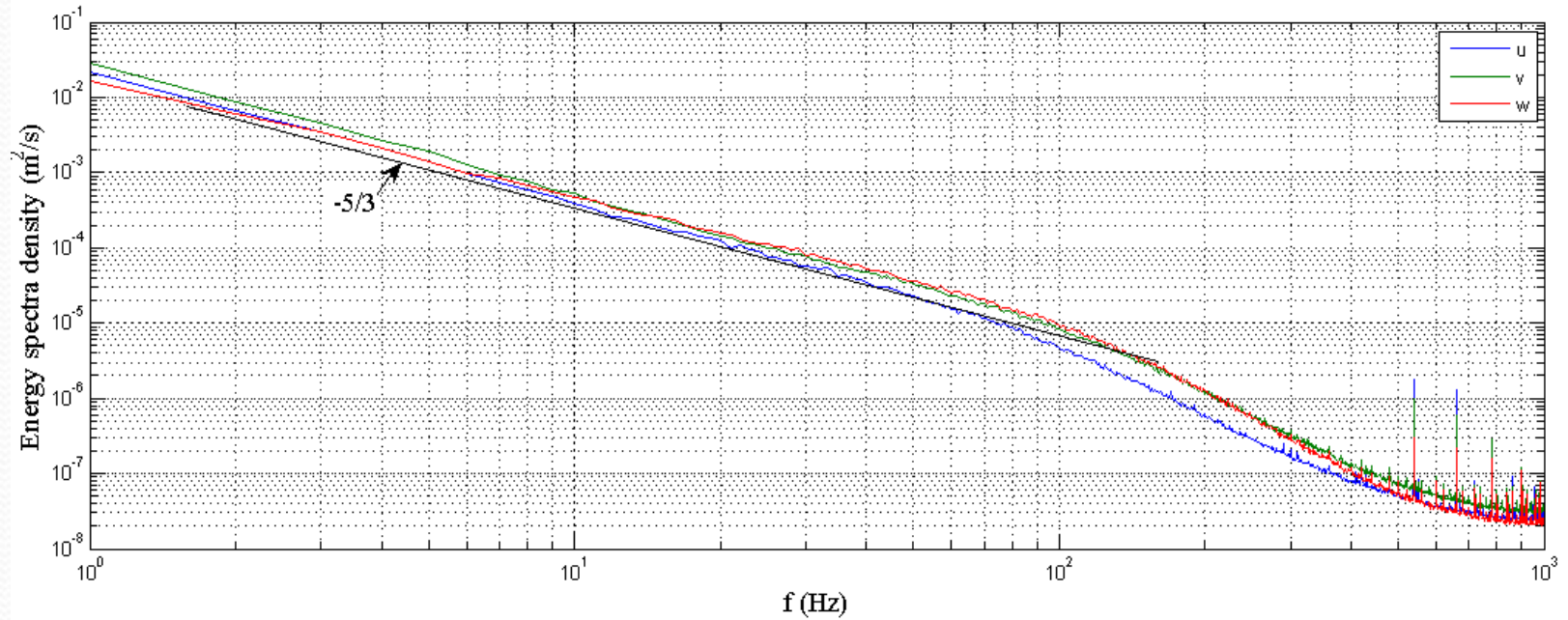
Burst/No-Burst Flag Selection



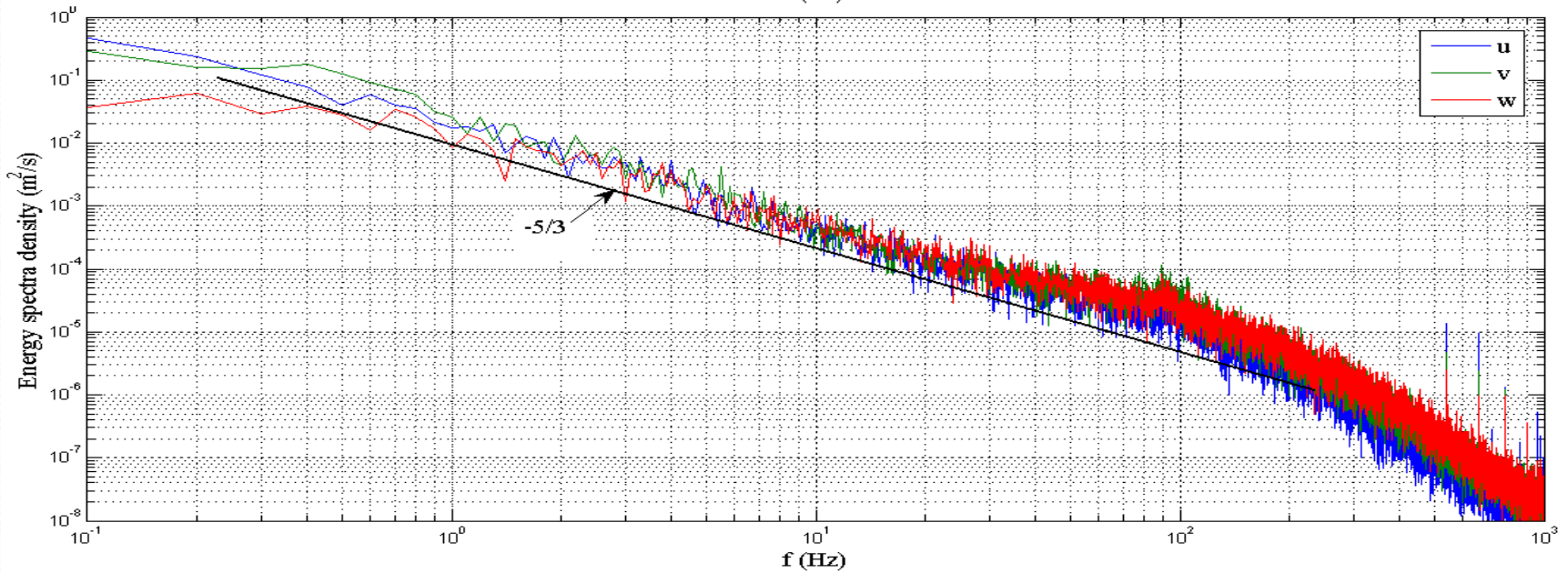
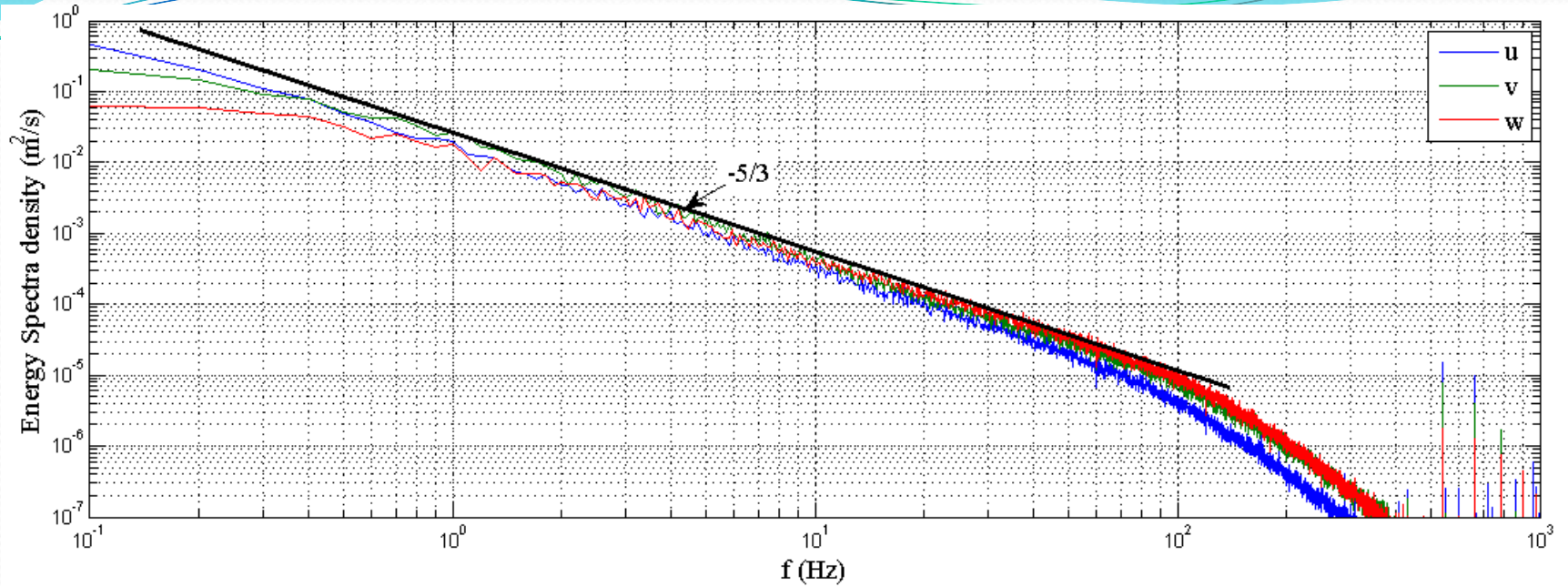
Velocity field at bursting events in forth S-I



SPECTRA: upper - no-burst, bottom - burst

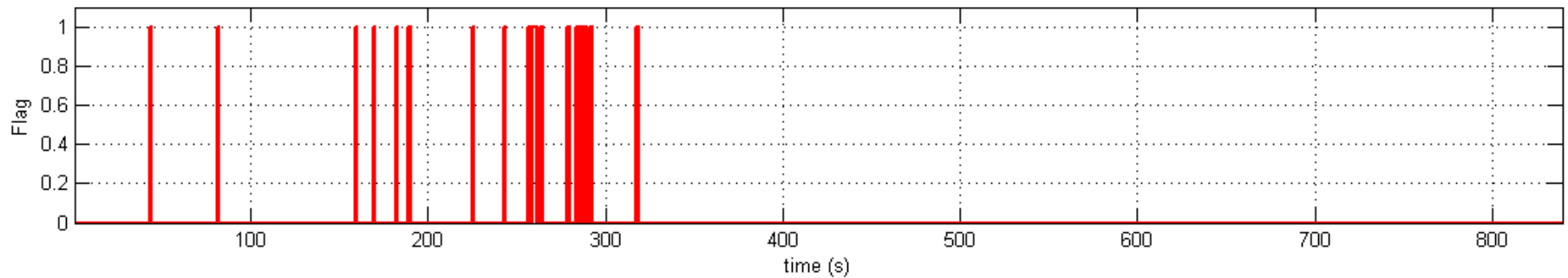
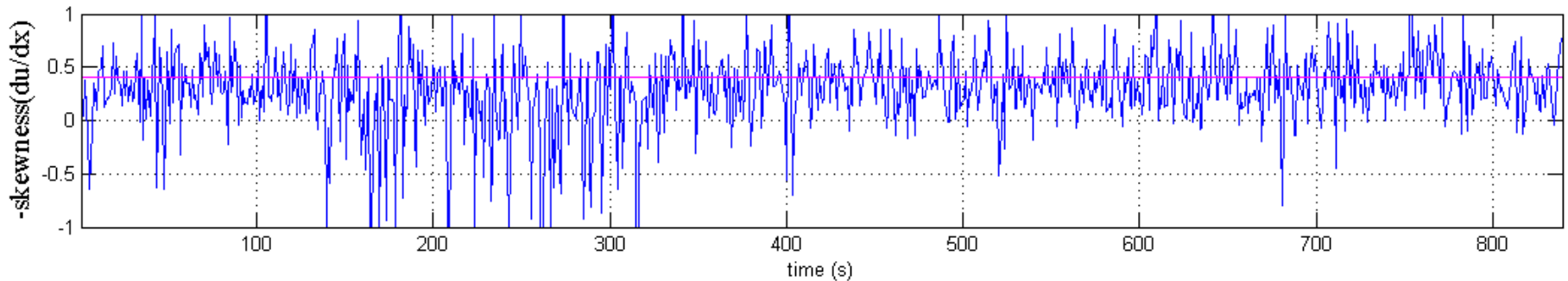


SPECTRA: upper - no-burst, bottom - burst

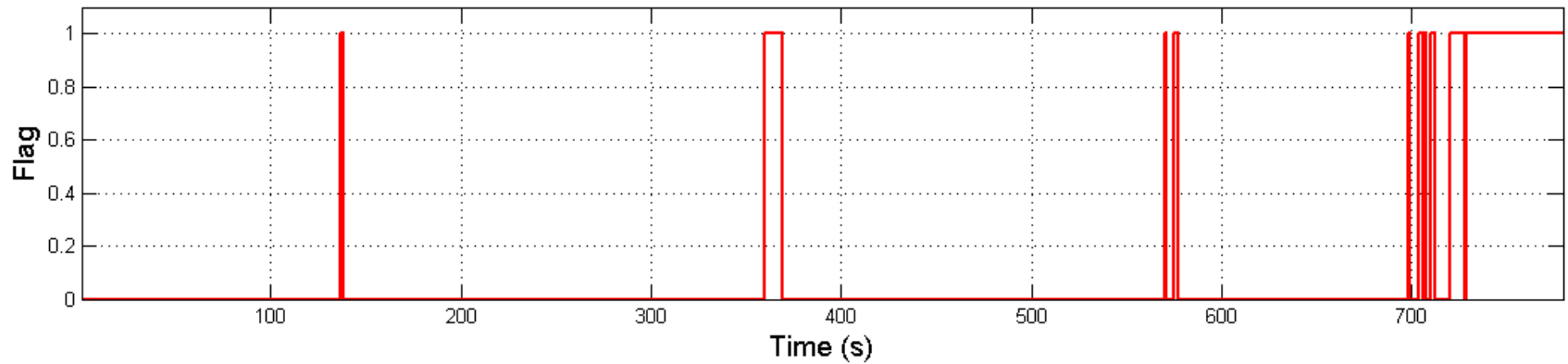
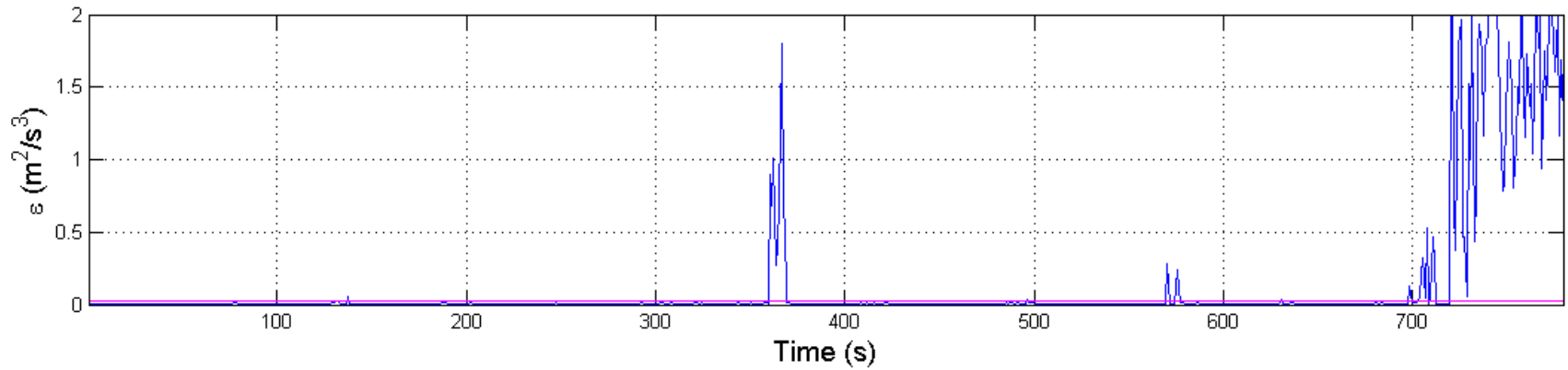


$$S = (\partial u / \partial x)^3 / ((\partial u / \partial x)^2)^{3/2}$$

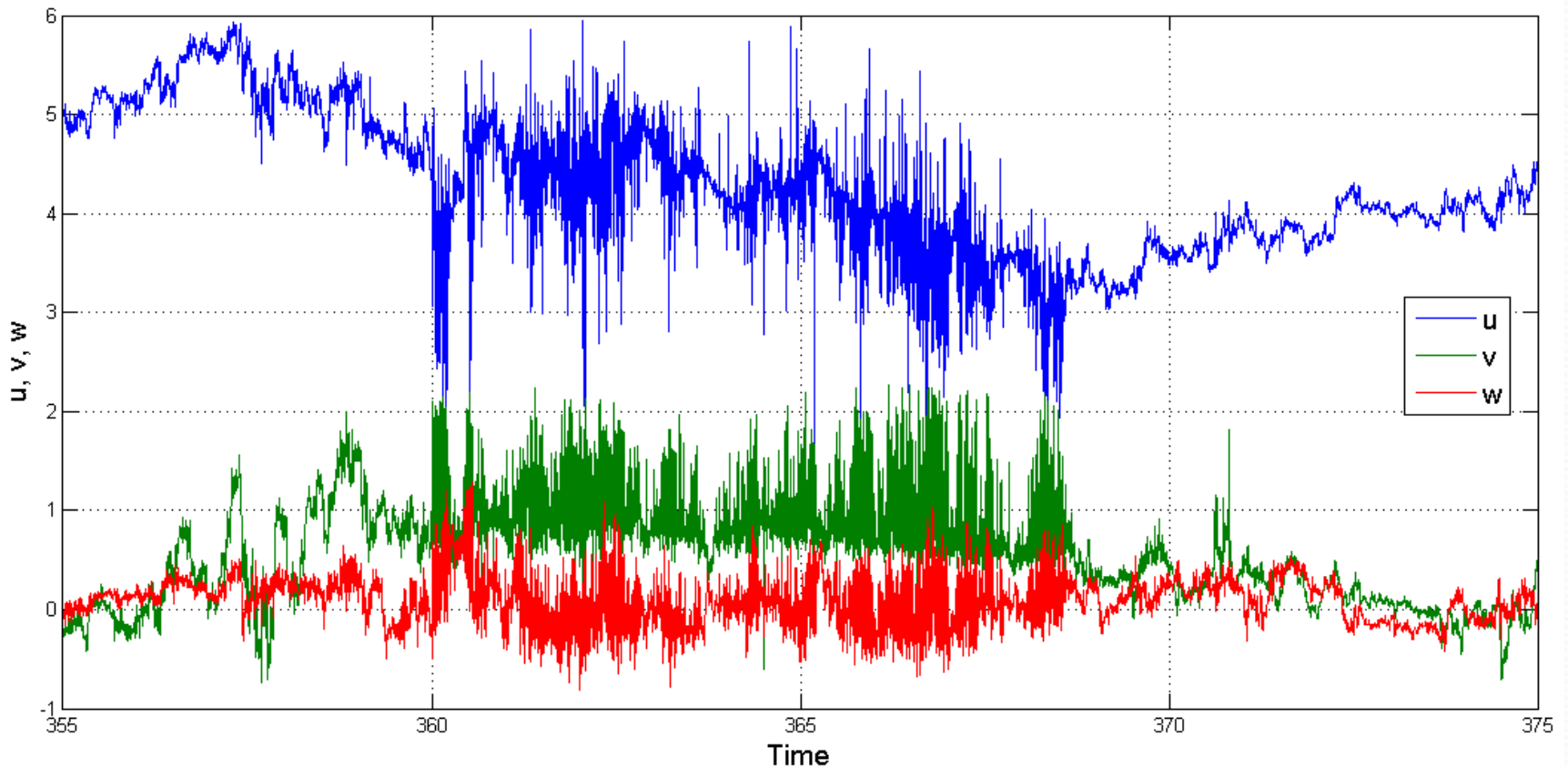
Times series of skewness of the longitudinal velocity derivative



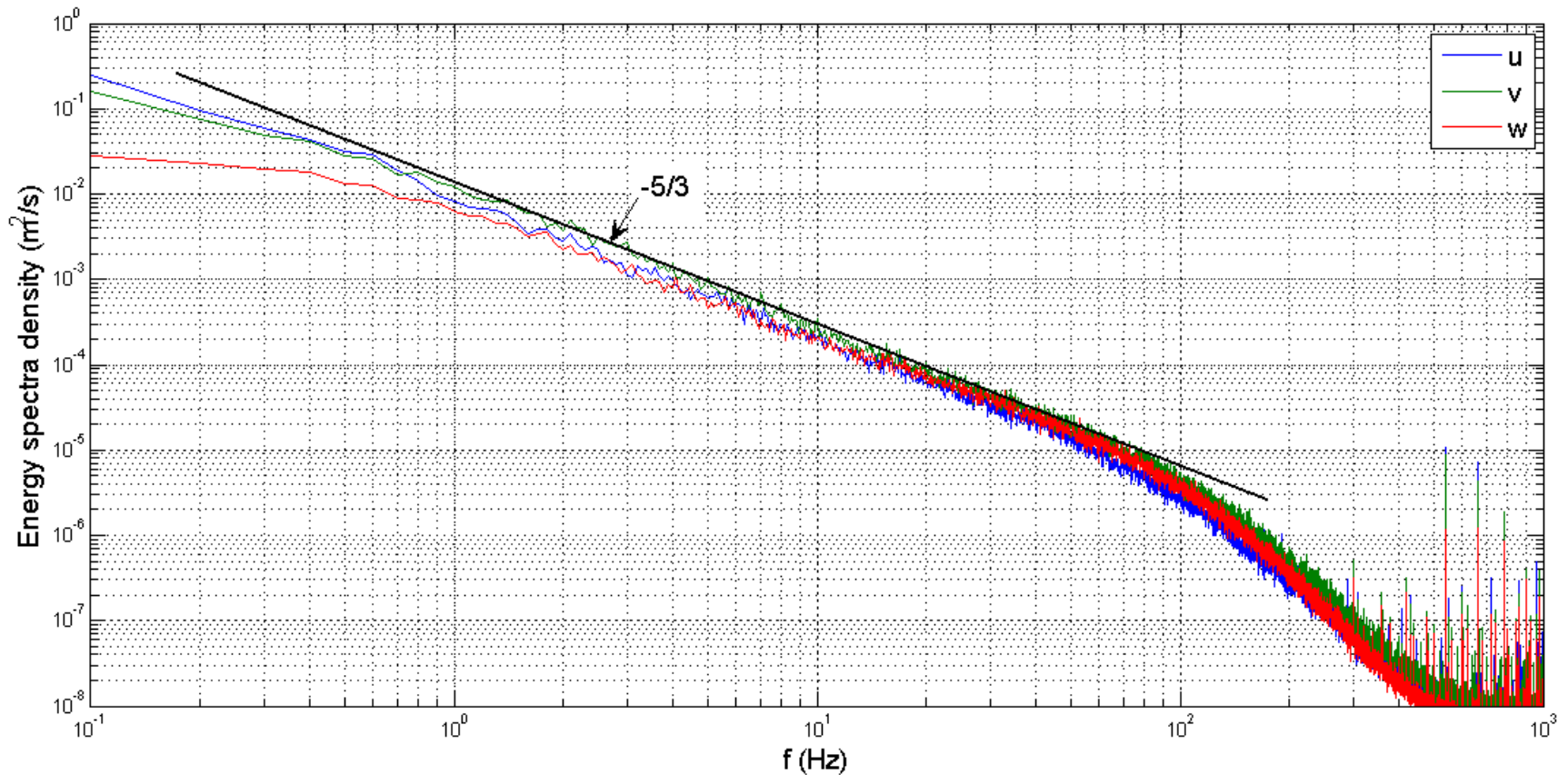
TKE dissipation at 5th S-I and Flag Selection



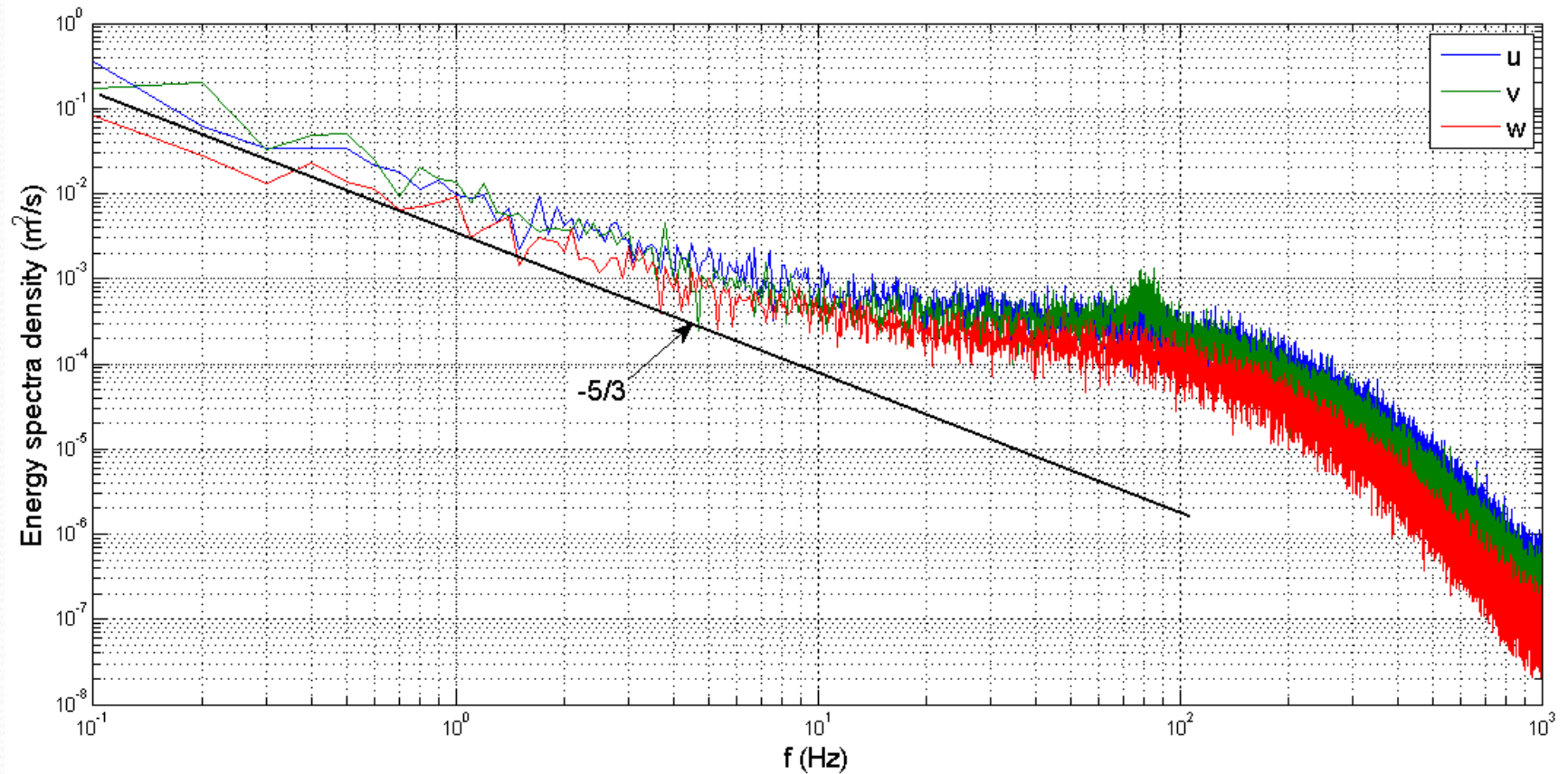
Velocity field at bursting events in fifth S-I



Kolmogorov spectra for non-busting events



Spectra at busting events



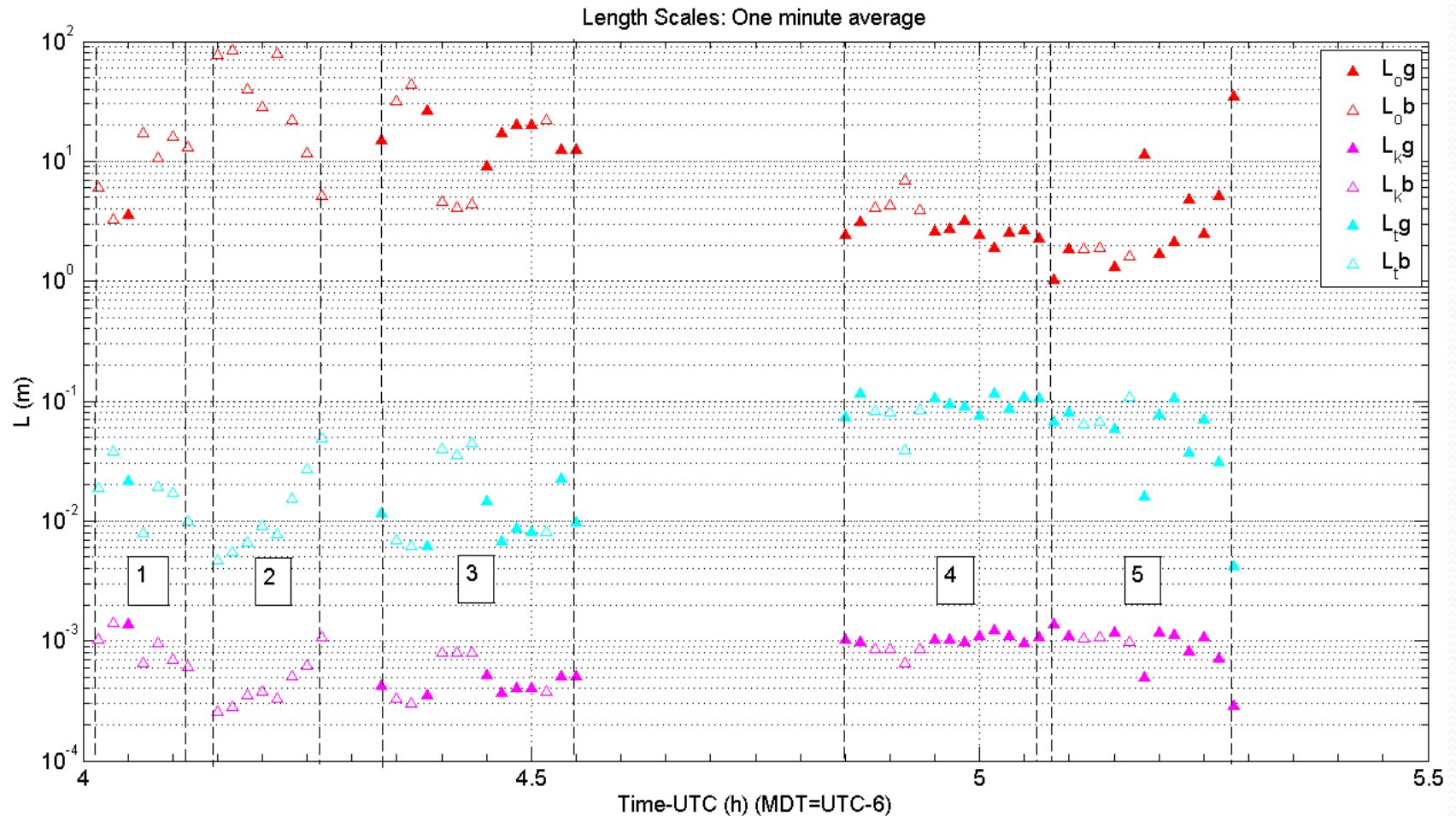
Mean and RMS velocities along with normalized Reynolds stress

Sub-intervals		Velocities	Time, (s)	U (m/s)	V (m/s)	W (m/s)	σ_u (m/s)	σ_v (m/s)	s_w (m/s)	coruw	Background ε_u (m ² /s ³)
1	No-burst		347	3.71	-0.053	0.034	0.070	0.075	0.050	-0.037	4.1e-4
	Burst		13	3.92	0.533	0.051	0.206	0.234	0.150	0.057	
2	No-burst		195	5.02	-0.201	0.035	0.200	0.187	0.179	-0.241	4.3e-3
	Burst		285	4.684	-0.046	0.006	0.394	0.314	0.303	-0.487	
3	No-burst		494	4.011	-0.238	0.045	0.176	0.192	0.166	-0.194	4.4e-3
	Burst		346	4.132	0.410	0.032	0.343	0.320	0.259	-0.362	
4	No-burst		816	4.338	0.031	0.046	0.207	0.237	0.199	-0.240	4.5e-5
	Burst		24	3.889	0.651	0.038	0.269	0.357	0.235	-0.269	
5	No-burst		702	3.936	0.037	0.005	0.158	0.175	0.133	-0.336	4.5e-3
	Burst		78	3.305	0.220	0.050	0.329	0.323	0.219	-0.257	

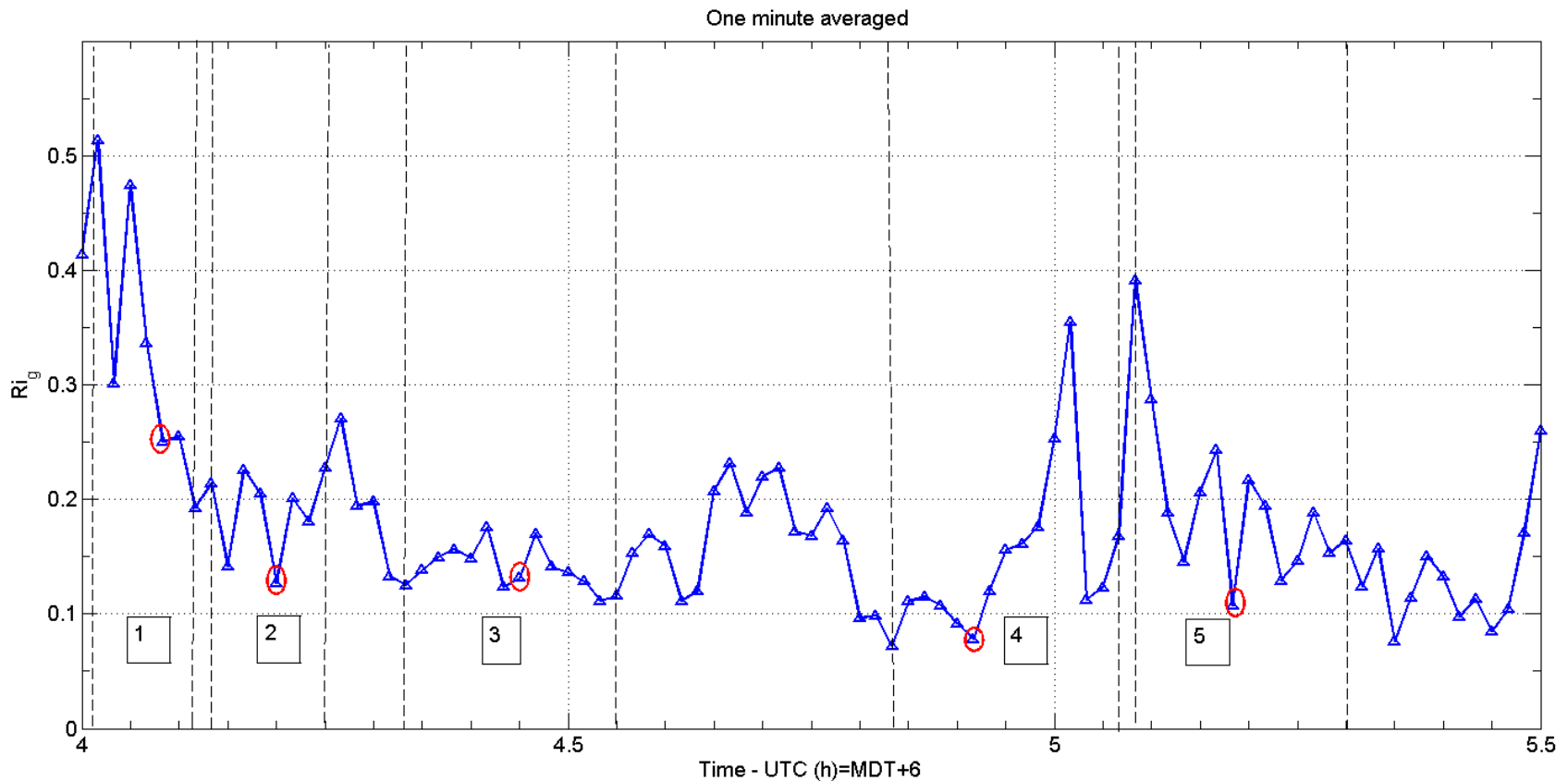
Characteristic lengths Scales

- **Ozmidov length scale** $L_0 = \sqrt{\varepsilon / N^3}$
- **Taylor microscale** $L_T = (\langle (u')^2 \rangle / \langle (\partial u' / \partial x)^2 \rangle)^{0.5}$
- **Kolmogorov microscale** $L_\eta = (\nu^3 / \varepsilon)^{1/4}$
- **Horizontal length scale** $L_h = \sigma_u^3 / \varepsilon$
- **Buoyancy length scale** $L_b = \sigma_w / N$
- **Shear length scale** $L_s = \sigma_u / (\partial U / \partial z)$
- ε is the TKE dissipation rate, ν is kinematic viscosity, N is the Brunt-Väisälä frequency
- (σ_u, σ_w) are the *rms* horizontal and vertical velocity

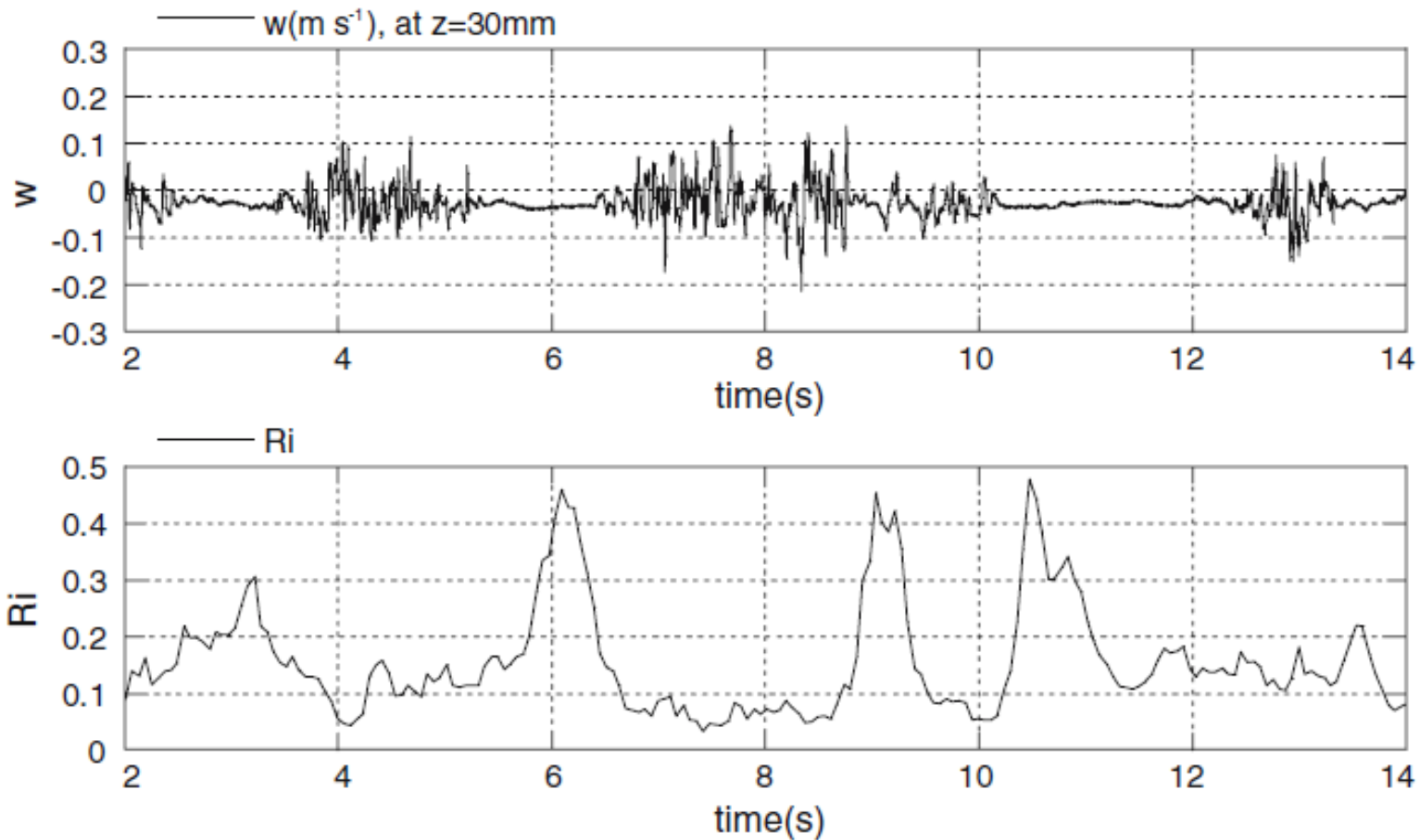
Time series of length scales during nocturnal transition. *(g)* represents “good” minutes judged by the value of the skewness S being in the range of $-0.2 > S > -0.6$ while *(b)* represents “bad” minutes whereas S outside of this range.



Rig vs. time during the evening transition. In red the selected minute when strong bursts of turbulence were observed. Numbers indicate the sub-intervals



Time histories of w velocity fluctuation and corresponding Richardson number at $z = 30\text{mm}$, CaseS2. Ohya et al. (2008)



Conclusions

- A combo system consisting of hot-film probes and collocated sonic was developed, with a capability of calibrating the hot-films using *sonic data and a Neural Network (NN)*.
- A platform was designed to automatically adjust the hot-film probes to the wind direction, prompted by in-situ sonic data.
- An automatic program to collect the data from sonic and hot-film probes enabled to obtain *continuous data* without interruption for more than 24 hours.
- In the present study *90 minutes of superb data* during an evening transition was measured from an almost laminar flow regime to a fully developed flow regime.
- A new phenomenon was discovered; *the occurrence of strong bursting of turbulence* related to the stable stratification.

Continuation...

- Processing procedures were developed for *separation of the turbulence to burst/no-burst events*.
- The spectra of events without bursting have *a classical Kolmogorov shape* while the events with bursting have a *bumping shape* resembling *bottleneck*.
- When the flow is quasi-stationary (e.g. 4th sub-interval), *skewness of velocity derivative* oscillates around **-0.4** for the events when the flow does not contain bursts.
- Various lengthscales are presented together with plots of energy spectral density. *The Ozmidov scale* separates regions strongly affected by stratification from the region where the classical isotropic Kolmogorov regime took place.
- The distribution of *the gradient Richardson number* was presented and qualitative correlation of the observed regimes with these Richardson numbers was suggested.

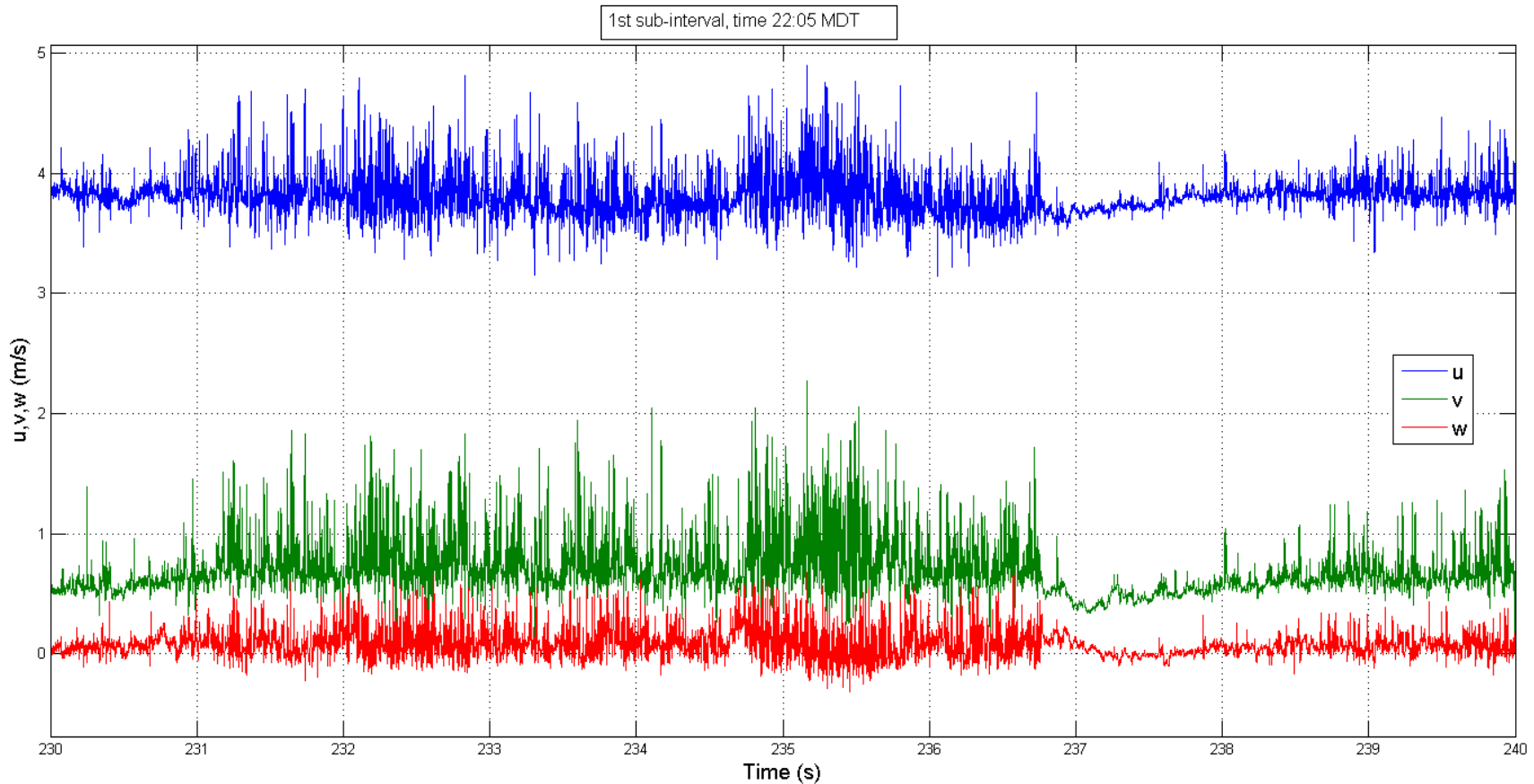
Continuation...

- Calculations of normalized correlation coefficient between longitudinal and vertical velocity components were performed separately for bursting and non-bursting events. *Similar values (in the range -0.4 to -0.2)* were obtained, thus supporting that bursts are related *to real physics* and not caused *by external excitation* due by noise.
- **There is a strong necessity to continue this research in order to clarify *whether bursting generation* was related to some specific *favorable conditions* in this complicated mountain terrain or represents *a generic phenomenon* that occurs due to strong stratification during evening periods. The answer to this question is of great importance for developing a more complete understanding of atmospheric boundary layers in mountainous terrain regions.**

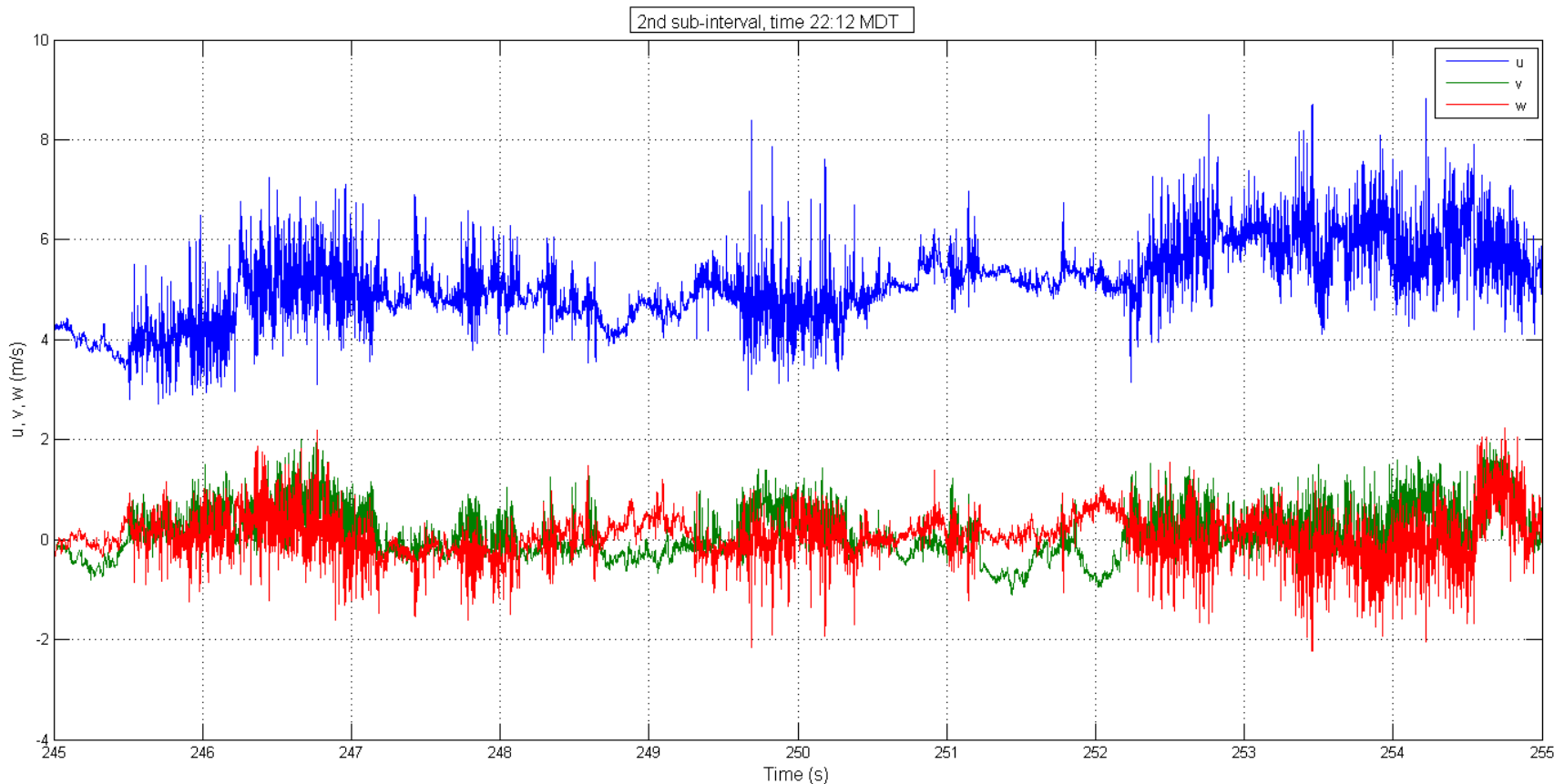


THANK YOU!
THE END

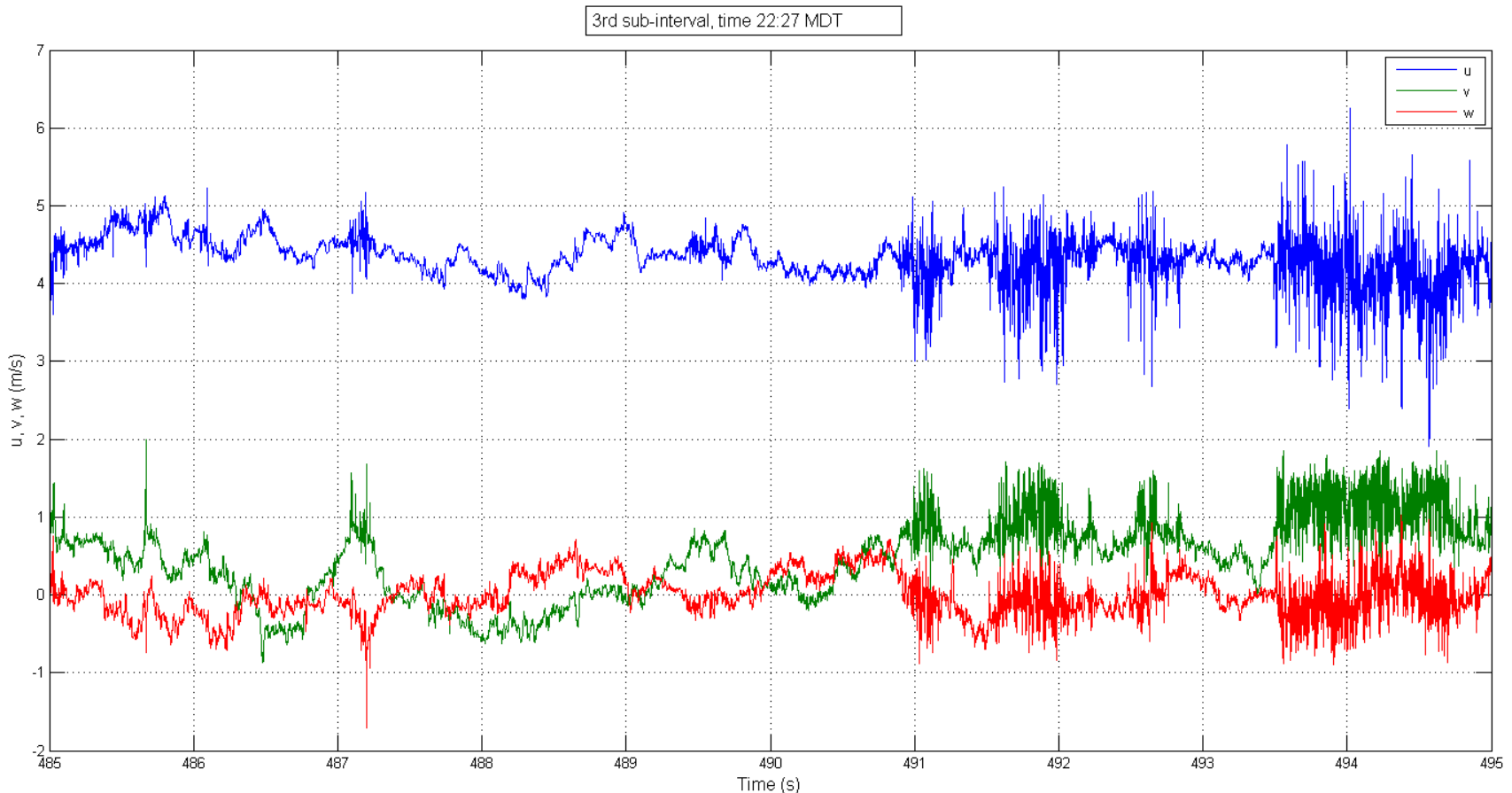
Three velocity components in a bursting event are shown for the end of fourth minute of the first sub-interval. Record duration of 10 sec is presented.



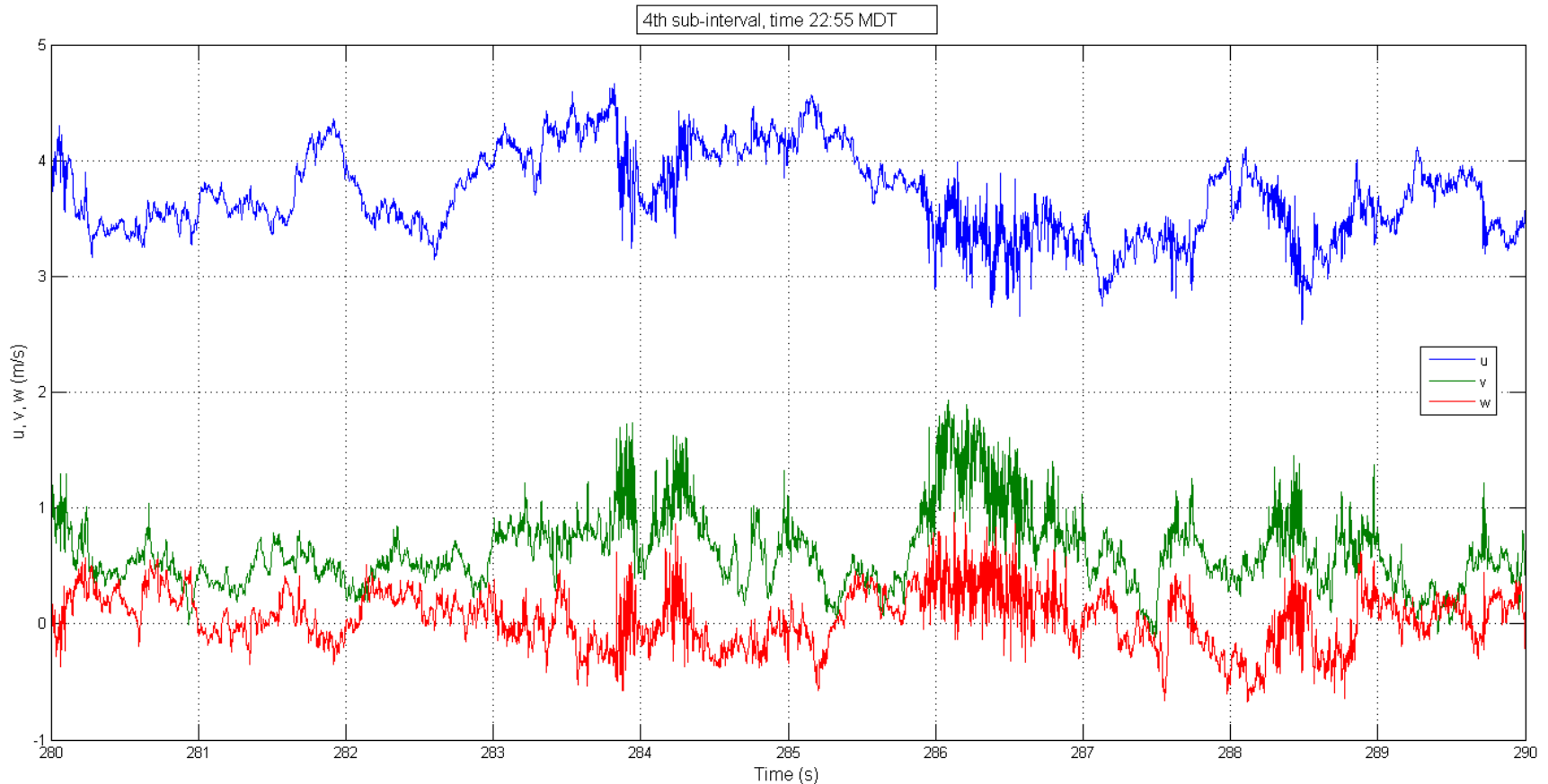
Three velocity components in a bursting event are shown for the beginning of fifth minute of the second sub-interval. Record duration of 10 sec is presented.



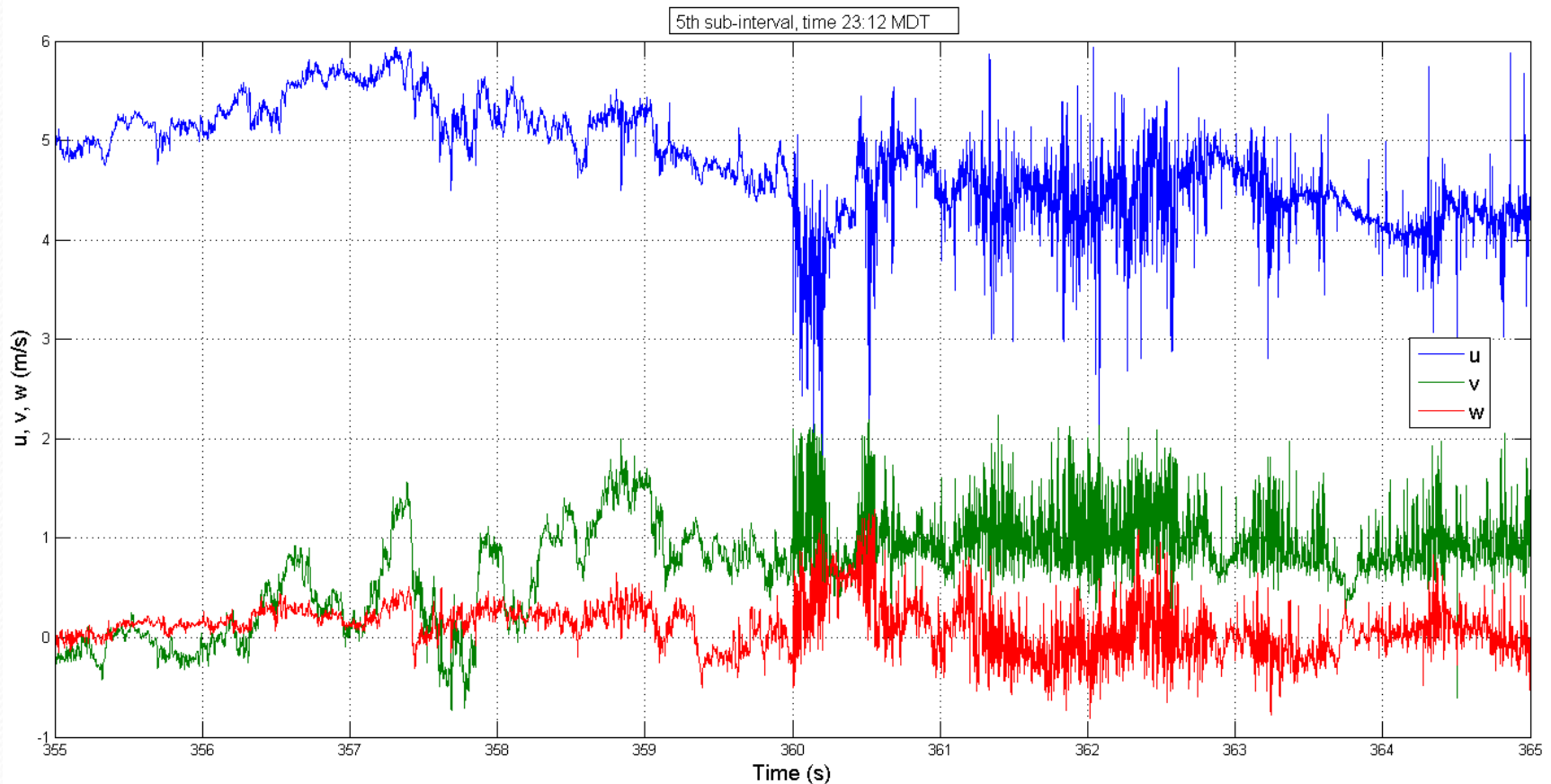
Three velocity components in a bursting event are shown for the beginning of sixth minute of the third sub-interval. Record duration of 10 sec is presented



Three velocity components in a bursting event are shown for the end of fifth minute of the fourth sub-interval. Record duration of 10 sec is presented

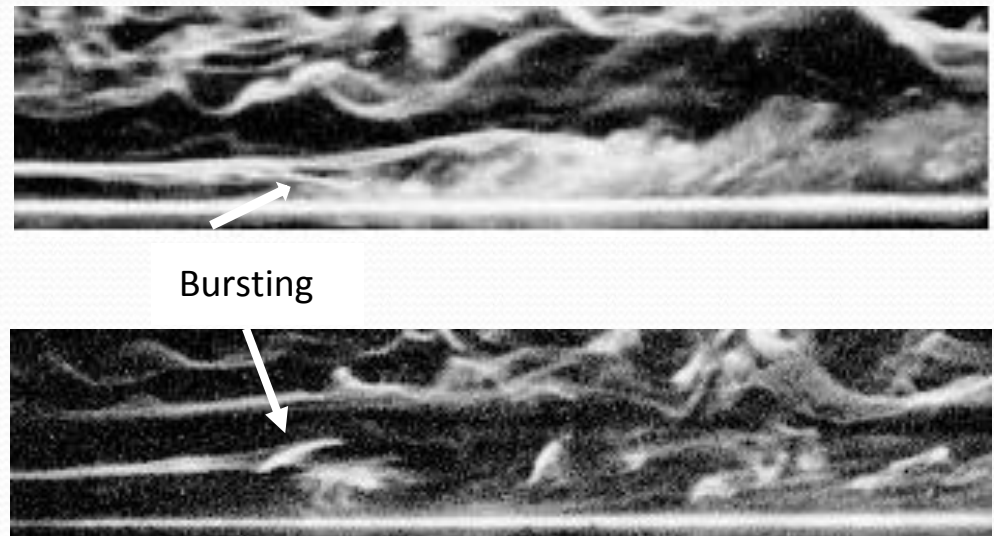
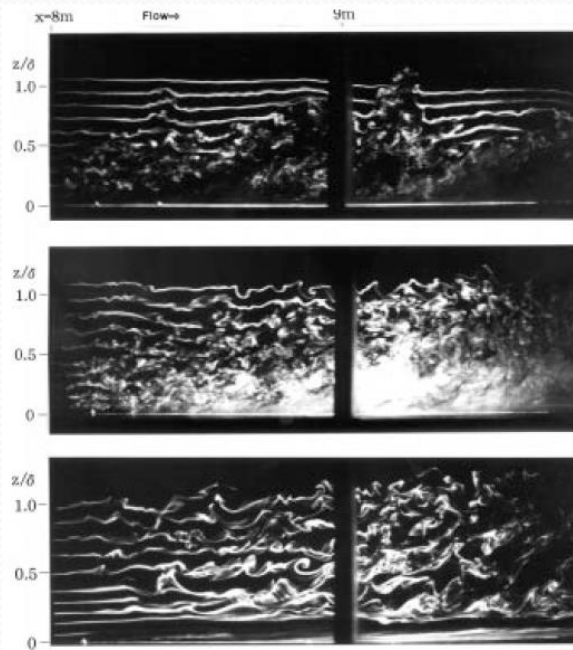


Three velocity components in a bursting event are shown for the end of sixth and beginning of seventh minutes of the fourth sub-interval. Record duration of 10 sec is presented

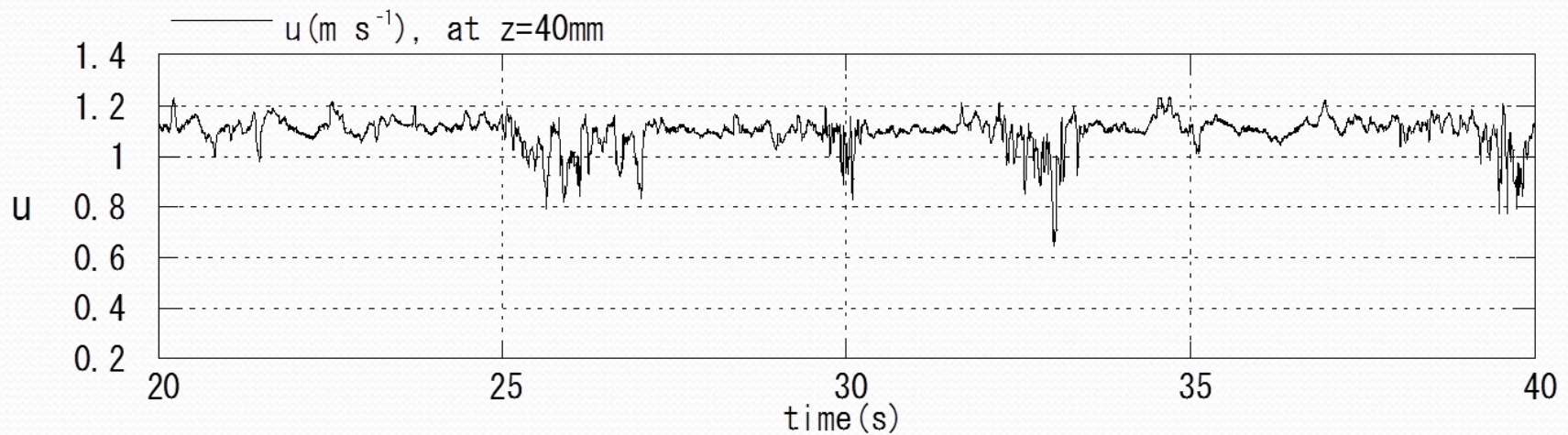


Flow visualization in laboratory experiments by Ohya (2001): top, middle and bottom plates are neutrally, weakly and strongly stratified BLs, respectively.

Flow visualization of SBL with a low-level jet, side view (Ohya et al. 2008, Case S2). Flow is left to right. Turbulent bursts and downward propagating turbulent puffs produced by a shear layer breakdown are clearly identified.



Time series of the longitudinal (u) velocity fluctuations in the laboratory experiments by Ohya et al. (2008), Case S2, showing a sequence of strong bursts



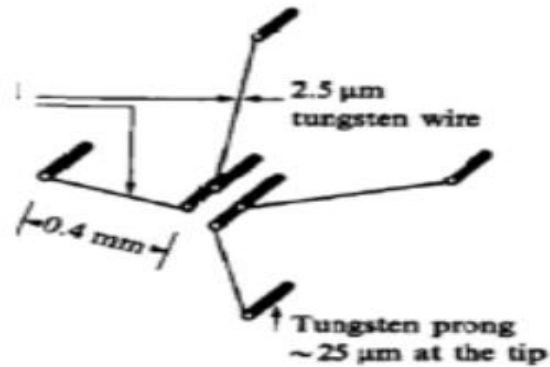


NEW HARDWARE DEVELOPMENTS

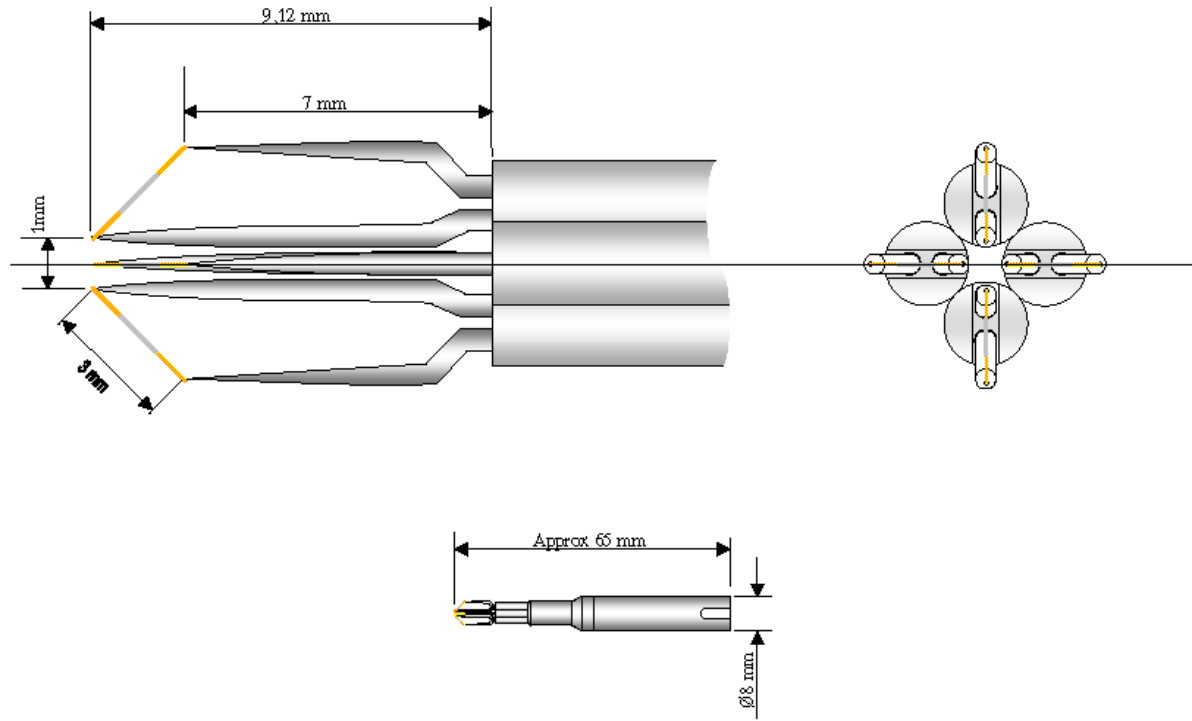
New supports for hit-film multi-sensors



4-wire array home-made and used by Tsinober, Kit and Dracos (JFM, 1992)



DANTEC Development of a new 3D-probe with 4 hot-film sensors.





Relevant Papers

- E. Kit, A. Cherkassky, T. Sant, H.J.S. Fernando. *In-situ* calibration of hot-film probes using a co-located sonic anemometer: Implementation of a neural network. **Journal of Atmospheric and Oceanic Technology-AMS, Vol. 27, No. 1, 23-41 (2010).**
- E. Kit and B. Gritz. *In-situ* calibration of hot-film probes using a co-located sonic anemometer: angular probability distribution properties. **Journal of Atmospheric and Oceanic Technology-AMS, Vol. 28, 104-110 (2011).**
- L. Vitkin, D. Liberzon, B. Gritz and E. Kit. Study of *in-situ* calibration performance of co-located multi-sensor Hot-Film and Sonic anemometers using a “virtual probe” algorithm. **Measurement Science and Technology 25, 075801 (2014)**
doi:10.1088/0957-0233/25/7/075801.

Presentation of velocity components as polynomials of voltages across the wires.

TKE dissipations and skewness of velocity derivatives

$$U_i = f_i(E_1, E_2)$$

$$f_i(E_1, E_2) = \sum_{kl} c_{ikl} P_k(E_1) P_l(E_2); \quad P_k(E) = E^k, \quad 0 \leq k, l \leq 4, \quad k + l \leq 4$$

Linear system for determination of polynomial coefficients c is obtained from calibration data using the least square fit.

Dissipation: $\epsilon = 15\nu \overline{\left(\frac{\partial u}{\partial x}\right)^2}$; $\partial x = -U\partial t$

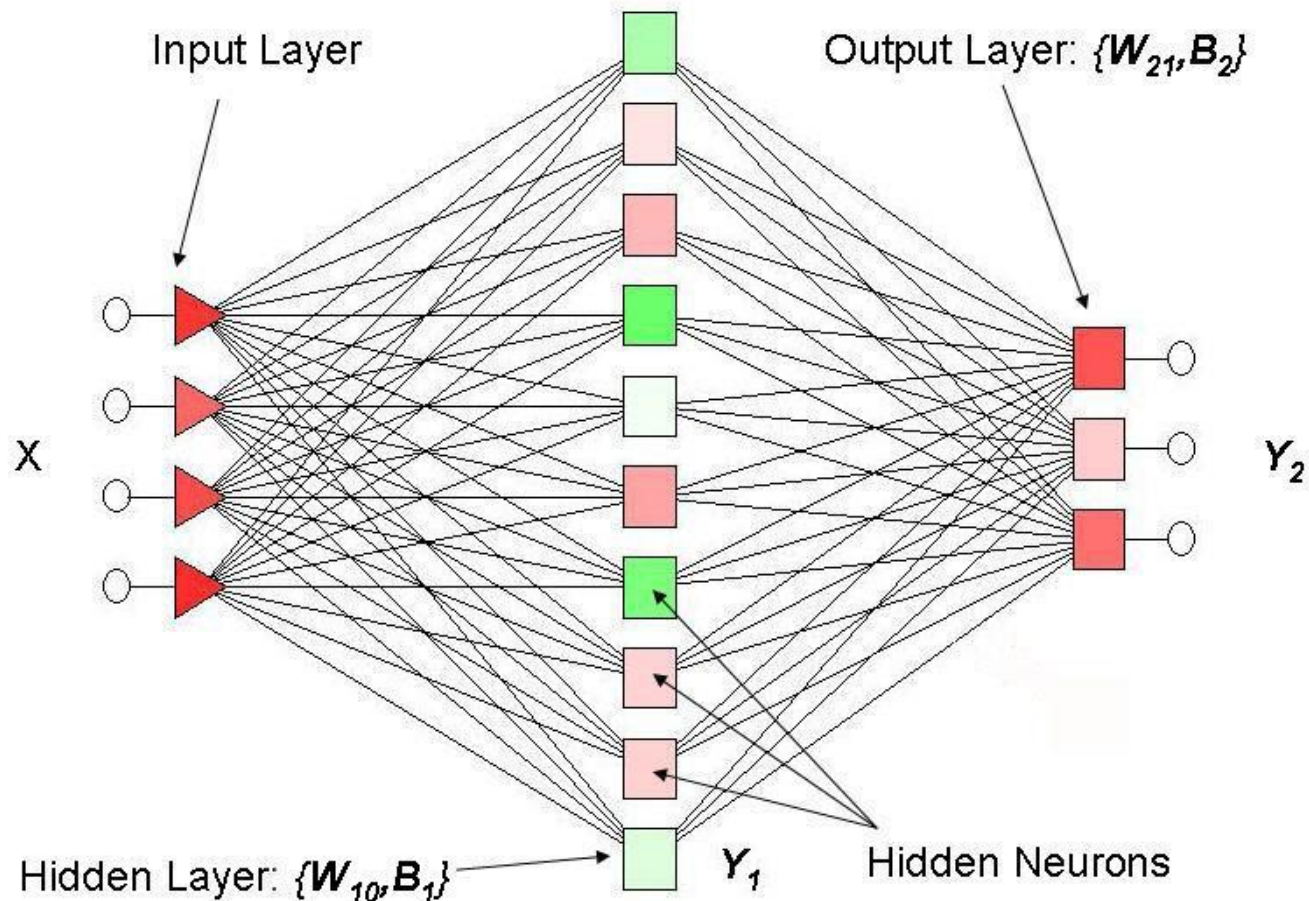
Skewness of velocity derivative: $Sk = \overline{\left(\frac{\partial u}{\partial x}\right)^3} / \left(\overline{\left(\frac{\partial u}{\partial x}\right)^2}\right)^{3/2}$

Combo in the Lab

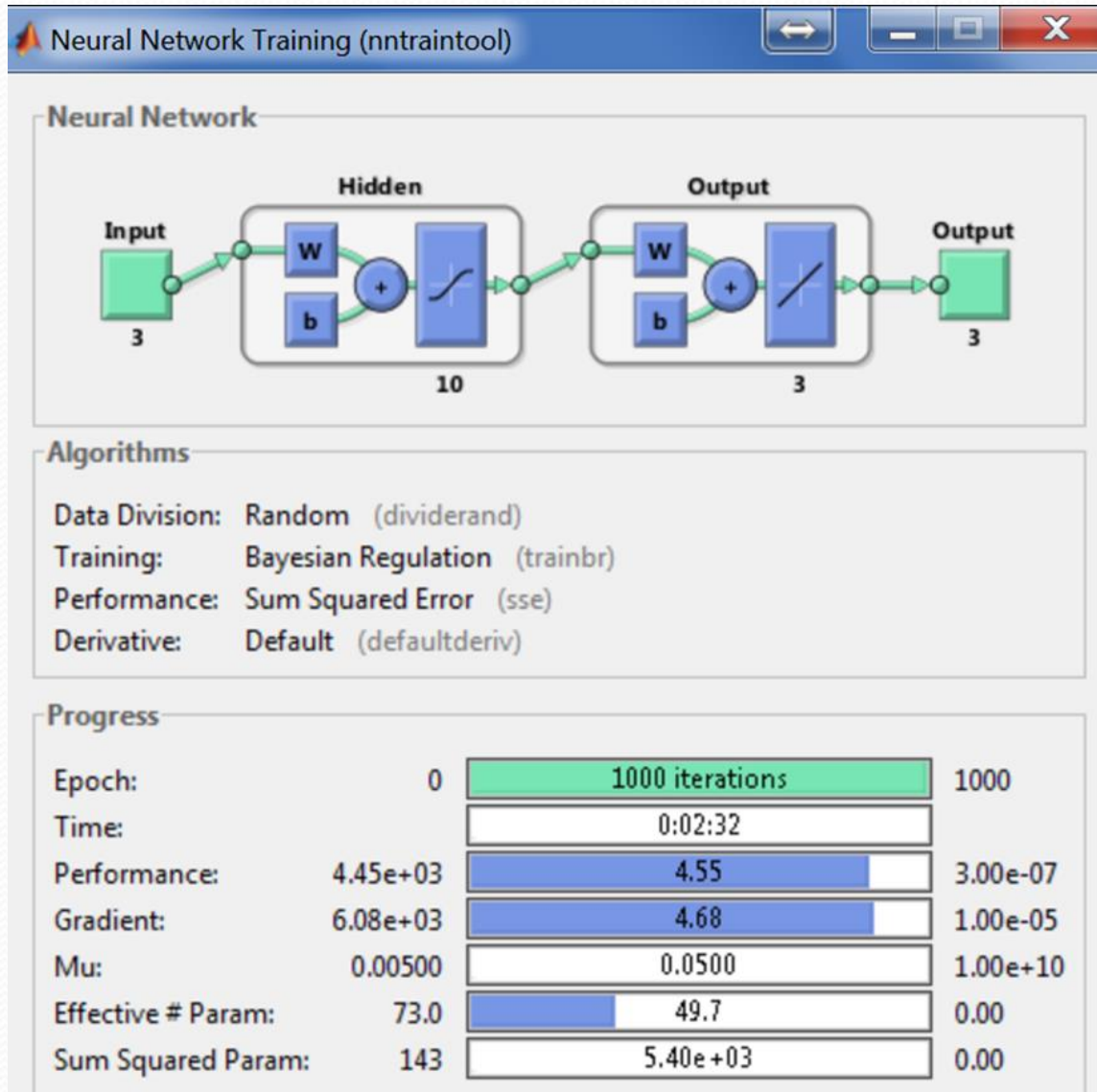


Neural Network

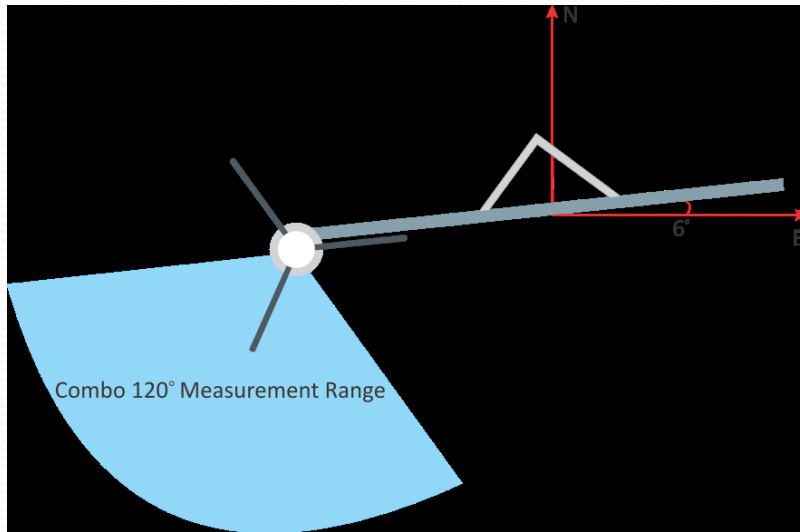
The structure of the generated neural network (3-layer Perceptron)



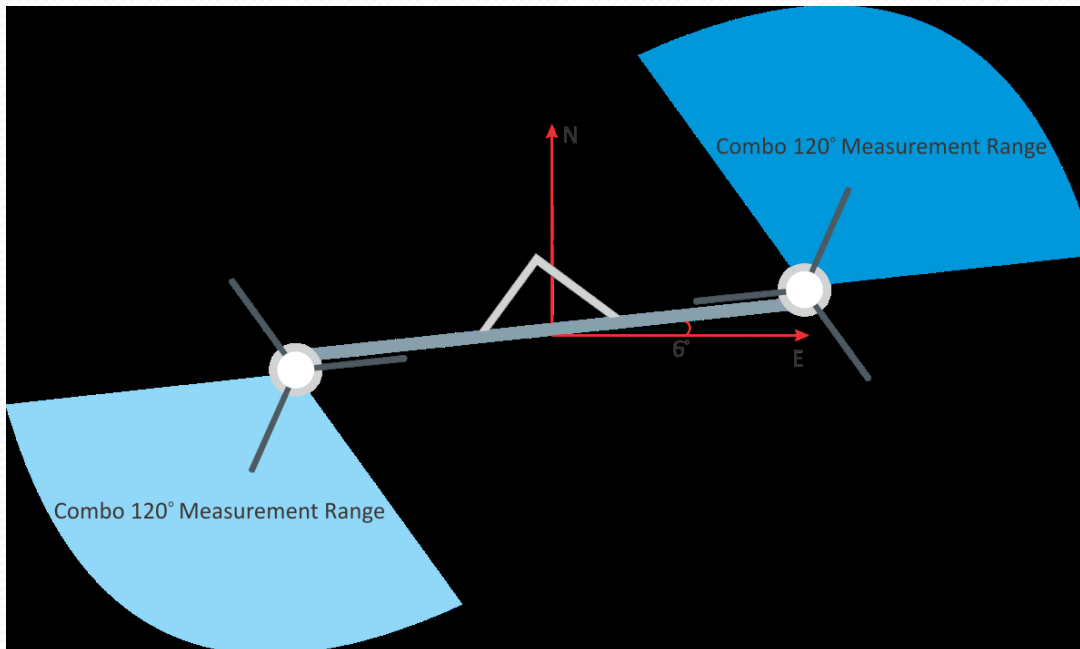
Neural Network Training using MATLAB



Combo probe Setup: (A) During the initial set-up (2&6m) (B) Second Half (both combos at 6m)



A



B

ES2 tower with combo probes positioned at 2 and 6 m.

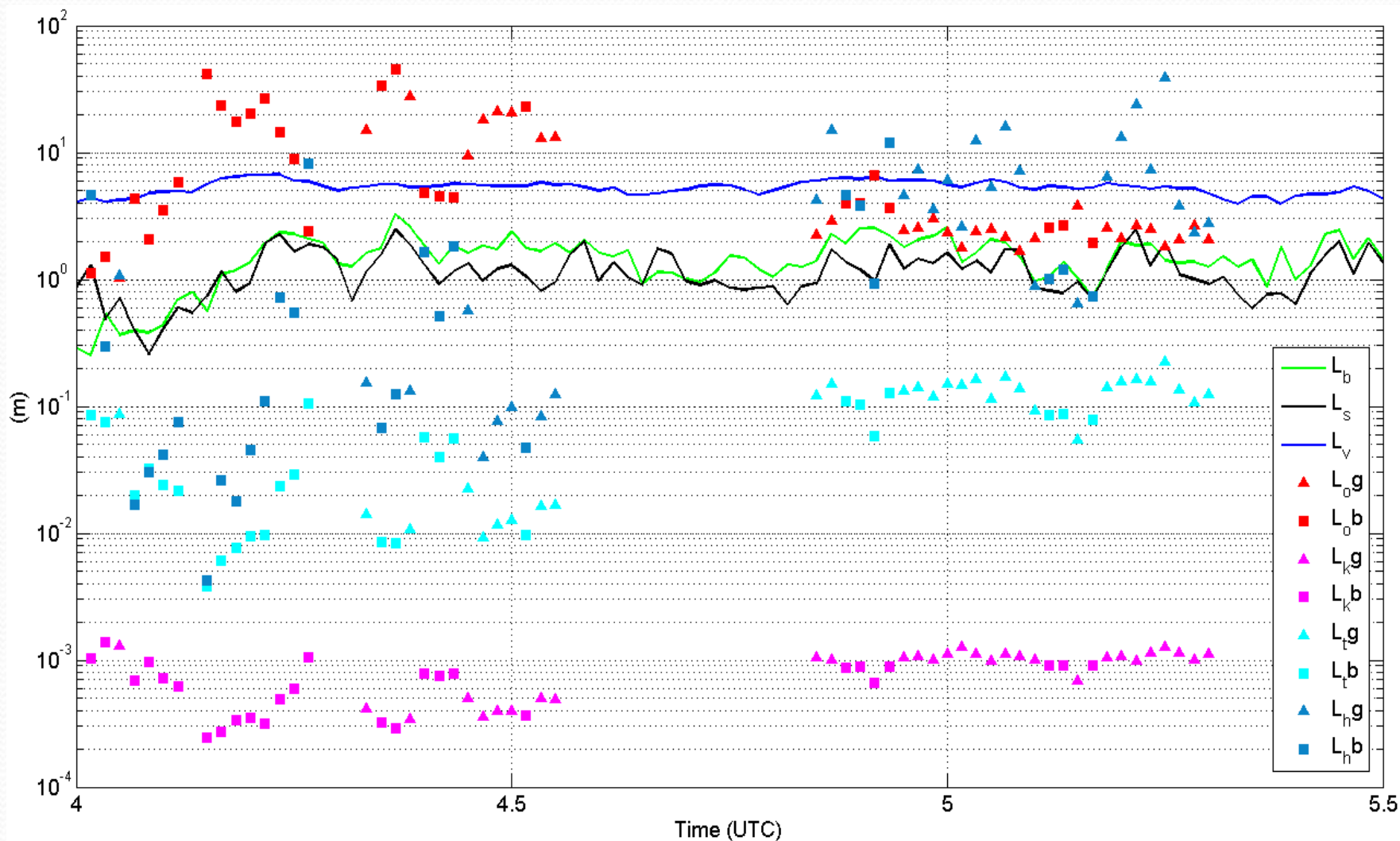


A

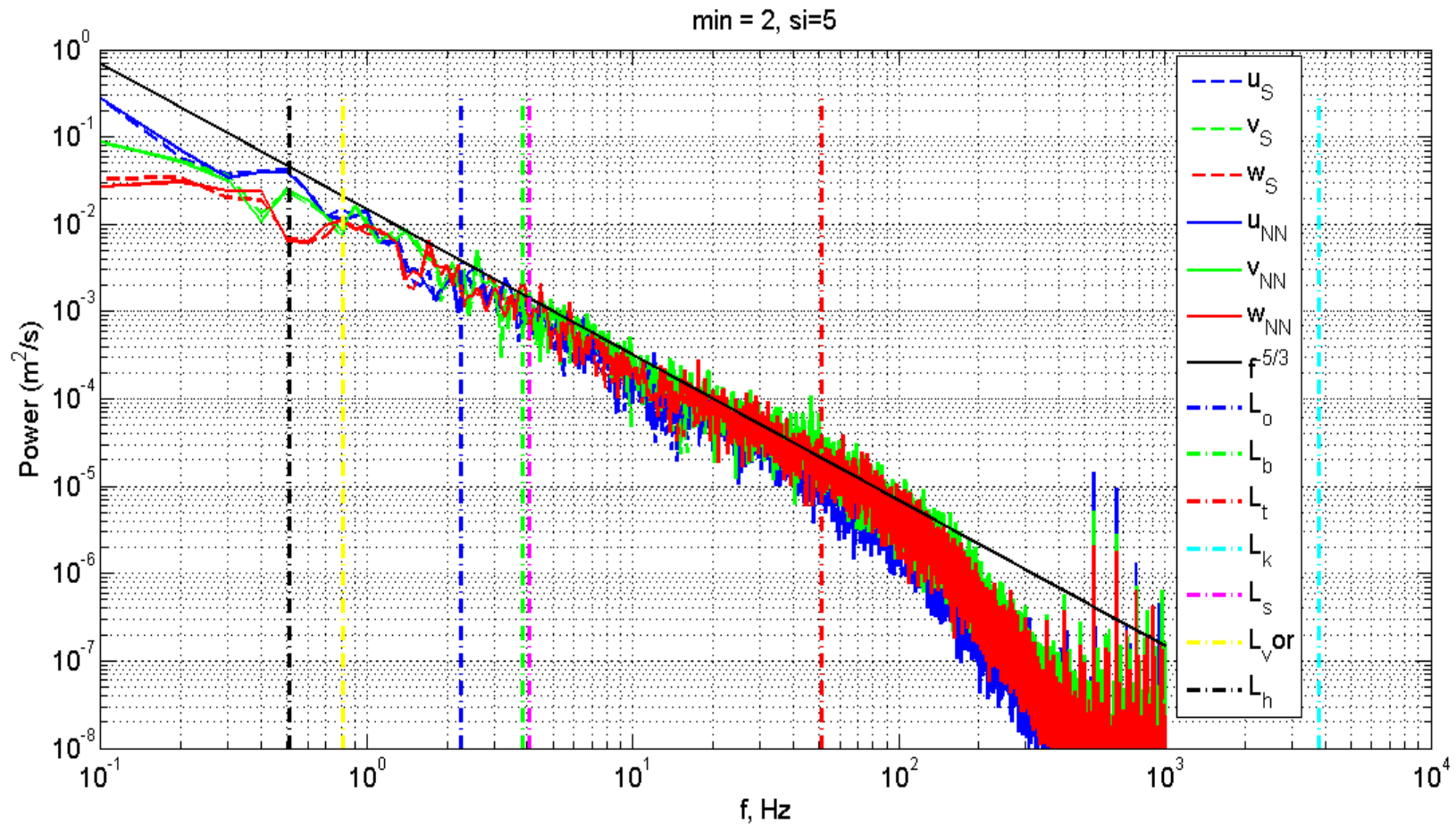


B ●

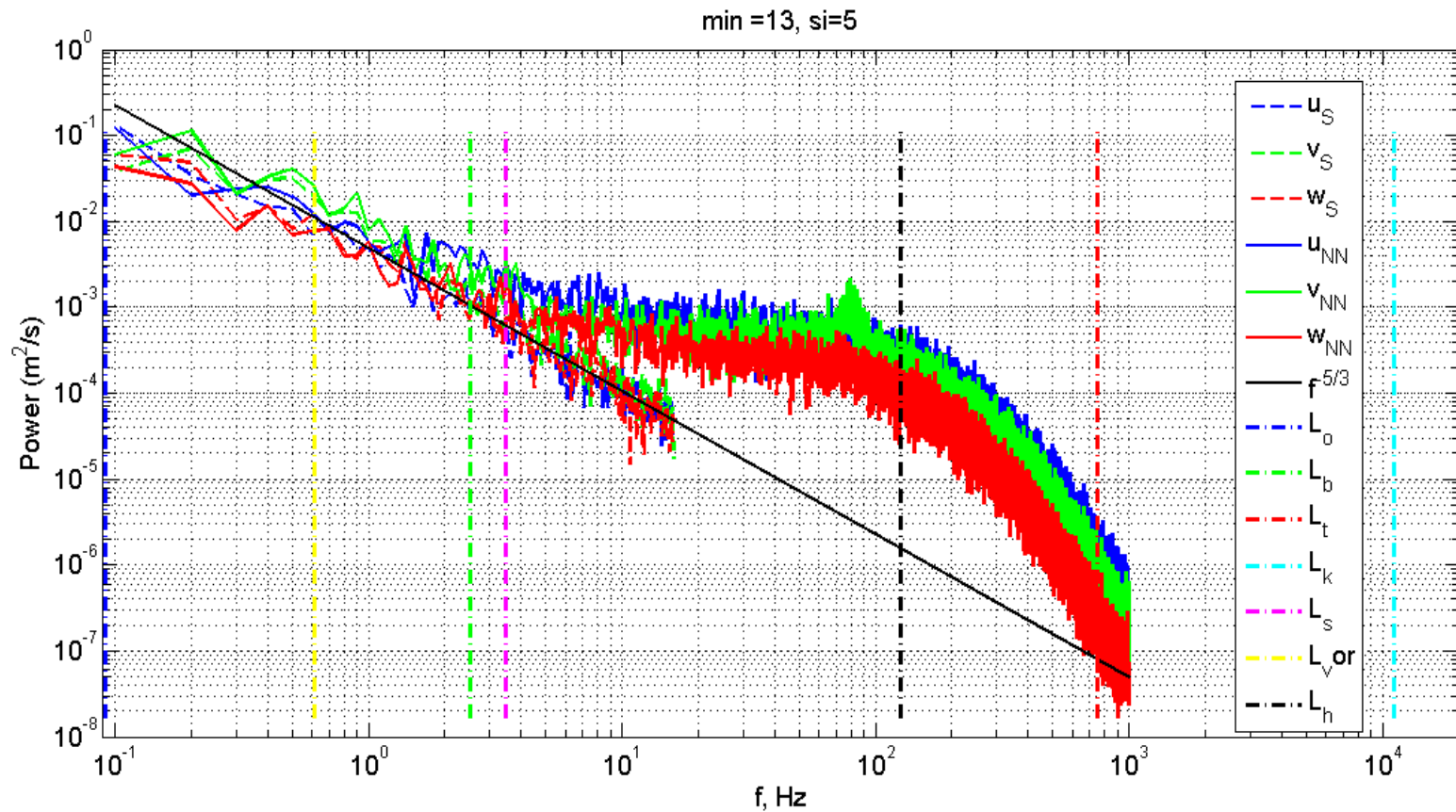
Characteristic lengths during evening transition



One-minute spectra and characteristic length scales



One-minute spectra and characteristic length scales





- **VIRTUAL PROBE ALGORITHM**

Virtual Probe

Determination of virtual probe parameters

- Calculation of the “effective” cooling velocity using calibration data-set previously measured (Kit et al., 2010)

$$U_{eff}^2 = U_n^2 + k^2 U_t^2$$

- Found best fit for King’s law coefficients **A**, **B** and the power **n**

$$E^2 = A + B \cdot U_{eff}^n$$

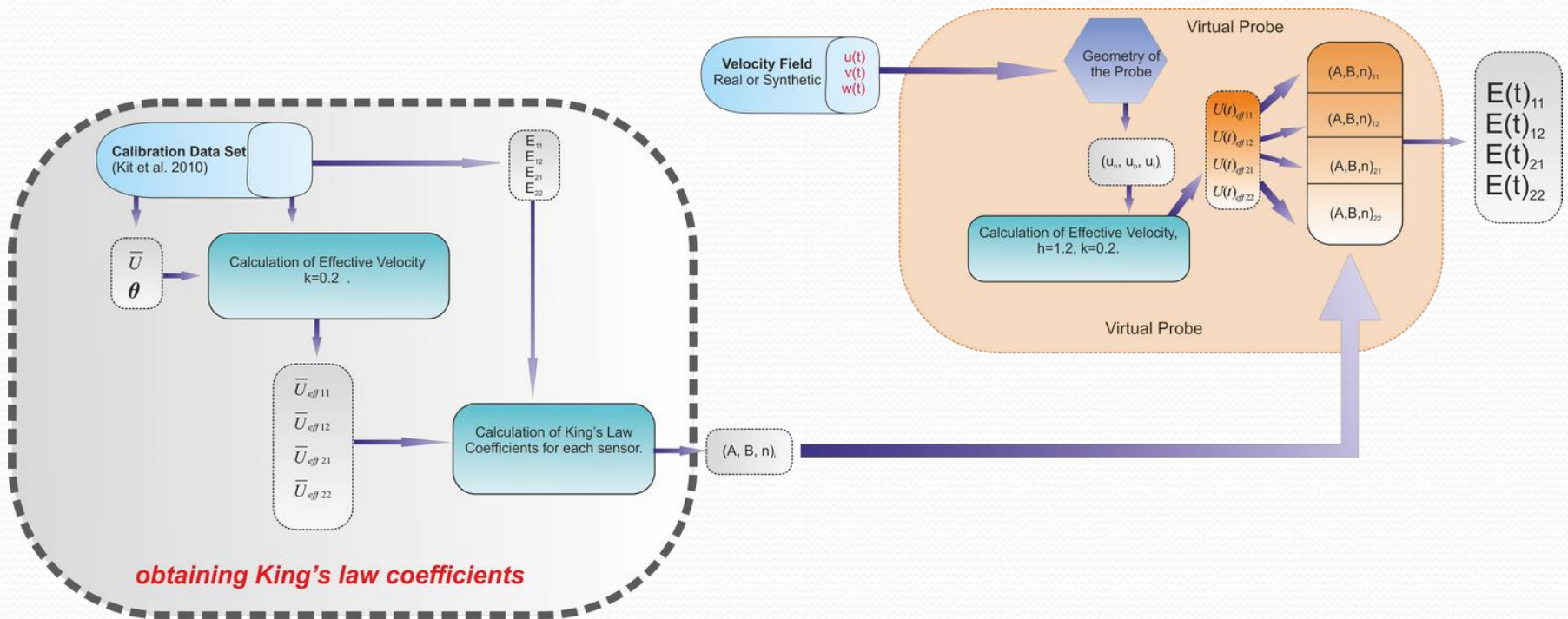
Computation of simulated voltages

- Calculation of the “effective” cooling velocity using measured or synthetic full velocity datasets

$$U_{eff}^2 = U_n^2 + k^2 U_t^2 + h^2 U_b^2$$

- Computation of simulated voltages employing King’s law with earlier determined coefficients **A**, **B** and **n**

Virtual Probe contd.



Conclusions

- Combo setup and Neural Network algorithm enable to obtain valuable information on the atmospheric flow especially during the transition events.
- Careful analysis is needed to select appropriate calibration datasets and time series for data processing
- The extrapolation of spectra based on Sonic data only can lead to a faulty conclusions as was indicated by velocity spectra obtained from combo measurements.
- There is indication that the use of four-sensor probes may be of advantage and can improve the signal-to-noise ratio due to redundant information.
- Further analysis of the spring data is in progress and hopefully will provide new perception

Conclusions

- NN model works with calibration datasets with unevenly distributed data points, PF works only with evenly.
- Field: Nocturnal works best and recommended.
- Very interesting spectra in our short preliminary campaign.
- Model of Angular Density Probability (ADP) is developed based on Gaussian distribution of velocity components.
- Angular Probability Distribution for calibration dataset is twice as narrow as for full signal. PF fails, NN comes through.
- Studying of non-linearity defined as RMS to mean velocity ratio
- Further development of the method: establishing of criteria for data quality.