









A Quality Assessment Scheme for Long-term Eddy-Covariance Measurements

Matthias Mauder, Matthias Cuntz, Clemens Drüe, Alexander Graf, Corinna Rebmann, Marius Schmidt, Rainer Steinbrecher

Atmospheric Environmental Research KIT/IMK-IFU



www.kit.edu



Outline

Introduction

- Quality assessment algorithms
 - Tests on high-frequency data
 - Tests on statistics, flux calculation and corrections
 - Quantification of errors and uncertainty estimates
 - The quality assessment scheme
 - Test data sets
- Results and discussion
 - Quality flags
 - Random errors and noise
 - Systematic error and energy balance ratio
 - Footprint analysis
 - Further tests
 - Effectiveness of the quality assessment scheme
- Conclusions

Introduction – Eddy Covariance



• Objective:

to measure the transport/flux of air properties (e.g. temperature) or air constituents (e.g. GHGs) between an ecosystem at the earth's surface and the atmosphere

Advantages:

- non-destructive
- no disturbance of the exchange conditions
- area-averaging on the ecosystem-scale
- quasi-continuous

Assumptions (Limitations):

- Atmospheric exchange is fully turbulent, zero mean vertical wind
- Taylor hypothesis: space and time interchangeable
- Horizontal homogeneity, stationarity/steady state conditions
- All transporting eddy scales are captured
- Constant flux layer

3

Introduction – Eddy Covariance



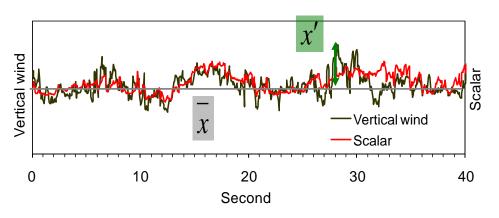


(Burba and Anderson, 2010)

Measurements at a single point can represent the ecosystem flux from an upwind area

Introduction – Eddy Covariance

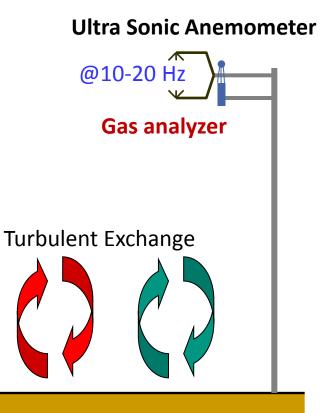




Flux Calculation (assume stationarity)

$$F = \overline{wq} = \overline{wq} + \overline{w'q'}$$

 $\overline{w} = 0$, assume horizontal homogeneity, 30 - 60 min averaging



Introduction – Challenges for QA scheme



- Extensive amount of data
 - Per site @ 20 Hz: 72000 data lines per hour, 1728000 lines per day, 51840000 lines per month, 1.89216 10¹⁰ lines per year
 - > 10 sites operational or planned within TERENO running for > 10 years
- Detection of instrumental failure
 - coverage by rain droplets, dew or frost
- Validity of assumptions for the EC method
 - well-developed turbulence, stationarity, zero mean vertical wind
- Quantification of measurement errors
 - noise, random, systematic

Representativeness of the flux estimate for the targeted ecosystem

footprint

Introduction – State of the art



- Foken and Wichura (1996, FW96), Foken et al. (2004)
 - Tests on stationarity, well-developed turbulence, zero mean vertical wind => flag system 1 – 9
 - Foken T, Göckede M, Mauder M, Mahrt L, Amiro BD, Munger JW (2004) Post-field data quality control. In: Handbook of Micrometeorology. A Guide for Surface Flux Measurements. Lee X, Massman WJ, Law BE, eds. Dordrecht: Kluwer, pp. 181-208.
 - Foken T, Wichura B (1996) Tools for quality assessment of surface-based flux measurements. Agric Forest Meteor 78:83-105.

Vickers and Mahrt (1997)

- Various tests on raw data => system of hard and soft flags
- Vickers D, Mahrt L (1997) Quality control and flux sampling problems for tower and aircraft data. J Atmos Oceanic Technol 14:512-526.

Papale et al. (2006)

- Test on fluxes without knowledge of raw data, plus uncertainty due to corrections
- Papale D, Reichstein M, Canfora E, Aubinet M, Bernhofer C, Longdoz B, Kutsch W, Rambal S, Valentini R, Vesala T, Yakir D (2006) Towards a standardized processing of Net Ecosystem Exchange measured with eddy covariance technique: algorithms and uncertainty estimation. Biogeosciences 3:571-583.

Error estimates by various authors

noise, random, systematic

- Lenschow DH, Mann J, Kristensen L (1994) How long is long enough when measuring fluxes and other turbulence statistics? J Atmos Oceanic Technol 11:661-673.
- Finkelstein PL, Sims PF (2001) Sampling error in eddy correlation flux measurements. J Geophys Res 106:3503-3509.
- Billesbach DP (2011) Estimating uncertainties in individual eddy covariance flux measurements: A comparison of methods and a proposed new method. Agric Forest Meteor 151:394-405.

Introduction – Goals



- Provide comparable flux data with simple quality flags and quantitative error estimates (random and systematic)
- Create a comprehensive quality assessment scheme of tests and algorithms, which can be applied automatically to long-term measurements (as strict as necessary and as simple/lean/efficient as possible)
- The selected tests and criteria should be as fundamental as possible to allow a wide applicability to different site conditions
- Provide pre-processed additional information about the fluxes in a standardised way, which may assist the user in deciding how to use and filter the data for his/her specific application.

Quality assessment algorithms



Quality assessment algorithms can be subdivided into three parts

- Tests on high-frequency data
- Tests on statistics, fluxes and corrections
- Quantification of error/uncertainty estimates

Algorithms - Tests on high-frequency data



- Modern micrometeorological instruments have internal quality tests and provide a diagnostic flag, e.g. Campbell CSAT3 or Licor LI-7500.
- Spike test based on Median Absolute Deviation (MAD), which is a robust measure for outlier or spike detection
- Limits for instrumental plausibility screening of the high-frequency data

High-frequency data points rejected by the above tests were replaced by an error code (NaN)

Alternatively for short data gaps: - repeat the last measured value

- linear interpolation

Algorithms – Tests on statistics



- Consensus about flux calculation and correction procedures is a pre-condition for the quality assessment scheme.
 - We follow the recommendations of Lee et al. (2004), which is in accordance with procedure used in the CarboEurope-IP software comparison of Mauder et al. (2008)
- Simple flag system, which follows the agreement of the 2nd CarboEurope-IP QA/QC workshop (Mauder and Foken, 2004):
 - Flag 0 high quality data, use in fundamental research possible
 - Flag 1 moderate quality data, no restrictions for use in long-term observation programs
 - Flag 2 low data quality, to be discarded

Lee, X., Massman, W., and Law, B. E. (2004) Handbook of Micrometeorology. A Guide for Surface Flux Measurement and Analysis. Dordrecht: Kluwer Academic Press, 250 pp.

Mauder M, Foken T, Clement R, Elbers JA, Eugster W, Grünwald T, Heusinkveld B, Kolle O (2008b) Quality control of CarboEurope flux data - Part 2: Intercomparison of eddy-covariance software. Biogeosciences 5:451-462.

Algorithms – Tests on statistics



- Minimum of 90% of the raw data available per averaging period
- Test the assumptions of the EC method (simplified after FW96):
 - Stationarity (>30% = flag 1; >75% = flag 2)
 - Well-developed turbulence (>30% then flag 1; >75% = flag 2)
 - Zero mean vertical wind velocity (>0.10 m s-1 then former *flag* +1; >0.15 then *flag* 2)
- Flux conversions and corrections cause an interdependence of the simultaneously measured flux estimates:
 - if flagλ E == 2 then former flagH + 1 else if flagH == 2 then former flagλ E + 1 else if flagλ E == 2 or flagH == 2 then former flagNEE + 1
- All these thresholds may appear somewhat arbitrary but are based on many years of experience of FW96, the authors of this work, and in part have been applied by >500 users of TK2/TK3 worldwide

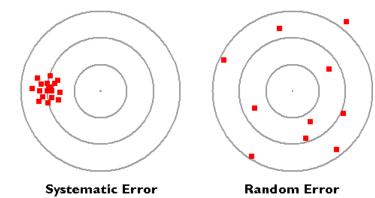


1. Instrumental noise

2. Random error

3. Systematic error





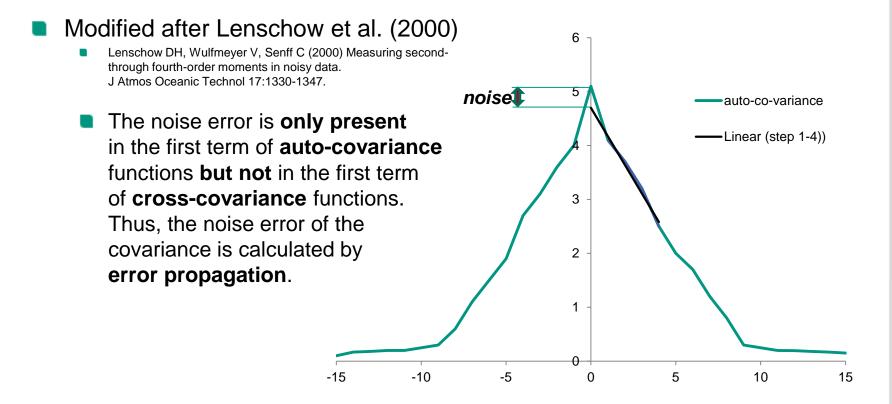
4. Representativity – Source area



Instrumental noise

14

- Spectral contribution in the high-frequency range
- Non-correlated within time series







Random error

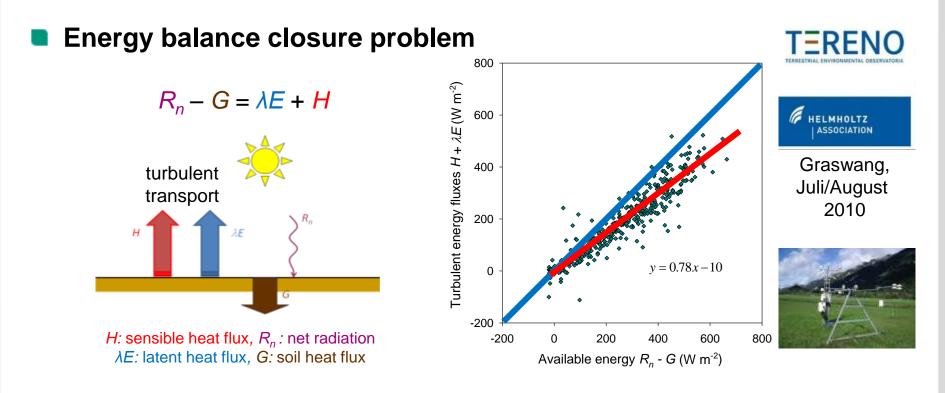
- Generally: ~ $1 / \sqrt{\# independent}$ observations
- Beware: # independent observations ≠ # observations, because the time series are auto- and cross-correlated.
- Modified after Finkelstein & Sims (2001)
- Finkelstein PL, Sims PF (2001) Sampling error in eddy correlation flux measurements. J Geophys Res 106:3503-3509.
 - The statistical variance of a covariance can be expressed as function of its auto-covariances and cross-covariance
 - An autocovariance is only meaningful for data without trend
 => detrending through high-pass filter before calculation of random error



Systematic error

- In the presence of large eddies, the covariance at a single point does not represent the total surface flux
- Why?
 - averaging interval too short
 - large eddies often "attached" to surface heterogeneities and do not propagate with the mean wind
- Capturing all relevant scales of biosphere-atmosphere exchange is subject of current research (among others: Helmholtz Young Investigators Group)
- Systematic error can only be determined indirectly



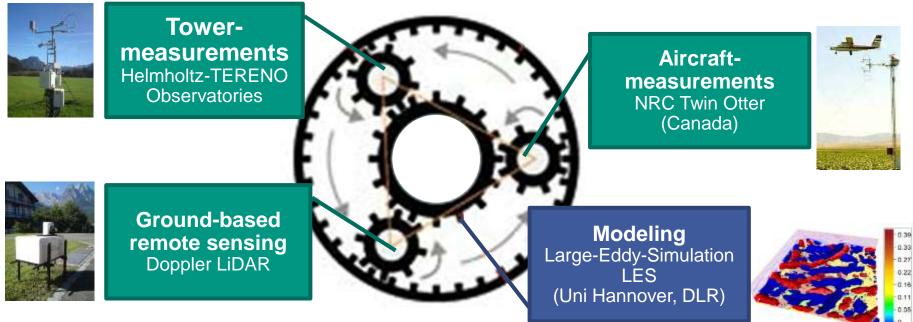


- In-situ measurements worldwide show:
 underestimation of the turbulent transport (λE + H) by 10-30%
- Strong evidence from 180 sites (Stoy and Mauder, 2011): transport on large scales (10² - 10⁴ m) not captured by measurements
- Stoy PC, Mauder M (2011) Energy balance closure at global-distributed eddy covariance research sites: the role of landscape-level heterogeneity. In: EGU General Assembly 2011. EGU2011-3066.

Helmholtz Young Investigators Group



Integrated research approach which interlinking modeling and measurements:



Goals

- Quantification/parameterization of the missing fluxes
- Improve the understanding of the effects of large-scale transport on flux measurements



Systematic error

Energy balance ratio EBR

$$EBR = \sum_{i=1}^{K} (H_i + \lambda E_i) / \sum_{i=1}^{K} (R_{n,i} - G_i - J_i)$$

• only applicable for daytime ($R_q > 20$ W m⁻²) and for 1-day averaging

the error is defined as

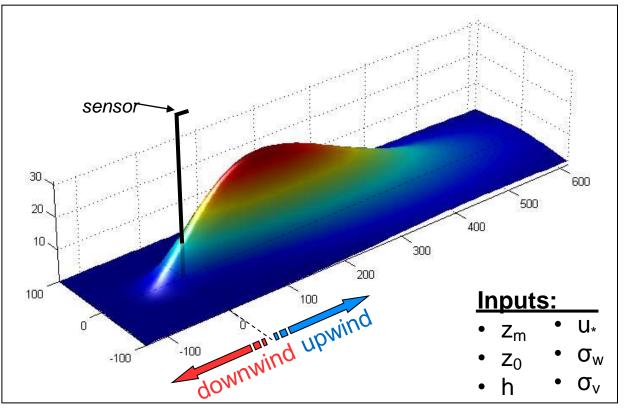
$$\sigma_F^{sys} = F \cdot \left(\frac{1}{EBR} - 1\right) \qquad R_g > 20 \text{ W m}^{-2}$$
$$\sigma_F^{sys} = n/a \qquad \qquad R_g \le 20 \text{ W m}^{-2}$$



Footprint = spatial filter, "field of view"

$$\mathsf{F}(\mathbf{x}) = \iint_{\Re} \mathsf{Q}_{s}(\mathbf{x}') \cdot f(\mathbf{x} - \mathbf{x}') \cdot d\mathbf{x}' = \mathsf{Q}_{s} * f$$

(convolution of the source distribution, Q_s , with the footprint, f)



Schmid HP (1994) Source areas for scalars and scalar fluxes. Bound - Layer Meteor 67:293-318.

Footprint-Model

Kormann R, Meixner FX (2001) An analytical footprint model for nonneutral stratification. Bound -Layer Meteor 99:207-224.

$$f = \frac{1}{\Gamma(\mu)} \frac{\xi^{\mu}}{x^{1+\mu}} e^{-\xi/x}$$

 $\xi(z) = \frac{Uz^r}{r^2\kappa}$, flux length scale

 $\mu = (1+m)/r$ constant

$$r = 2 + m - n$$
 shape factor

Analytical

- Numerically robust (log-profiles into power laws)
- Computationally fast and reliable enough to be applied for every 30-min interval in a long-term measurement programme



Karlsruher Institut für Technologie

The complete quality assessment scheme

Tests on high-frequency data

- Instrument diagnostic flags (CSAT3 0-63, LI7500 240-251)
- · Instrumental/plausibility limits (site-specific)
- Spike-detection with MAD-test, z = 7

Tests on statistics, flux calculation + corrections

- Maximum number of missing values: ≤ 10%: flag = 0, >10%: flag = 2
- Calculate averages, variances and covariances on 30 min and 5 min basis with automatic lag correction by maximising covariances
- Stationarity test covariances (FW96, < 30%: flag = 0, < 75%: flag = 1)
- Flux calculation, conversions and corrections applied: Planar fit,
- Schotanus, spectral losses (Moore or Eugster&Senn), WPL, iteration
- ITC test (FW96, MF04, < 30%: flag = 0, < 100%: flag = 1)
- Combination of flags according to MF04
- w after Planar fit > 0.10 m s⁻¹: former flag +1, > 0.15 m s⁻¹: flag = 2
- · Interdependence of flags due to corrections/conversions:
 - if $flag\lambda E == 2$ then former flagH + 1
 - else if flagH == 2 then former $flag\lambda E + 1$
 - else if $flag\lambda E == 2$ or flagH == 2 then former flagNEE + 1

Quantification of errors/uncertainty estimates

- Random error: use Finkelstein&Sims (2001), but on high-pass filtered (detrended) time series and on full period. Consider Gaussian error propagation for u-; take relative error of covariances and multiply with corrected fluxes (neglecting error propagation through corrections)
- · Noise error after Lenschow et al. (2000)
- Systematic error: Very low-frequency eddies and non-propagating eddies lead to a flux underestimation and consequent lack of energy balance closure, only applicable for R_g > 20 W m⁻² for one day; the error is flux*(1/EBR-1)
- Footprint: Kormann&Meixner (2001) using measured <u> instead of z₀dependent wind profile; calculate percentage of flux contribution for several targets of interest; recommended percentage threshold 70% depending on user's requirements

The test data sets



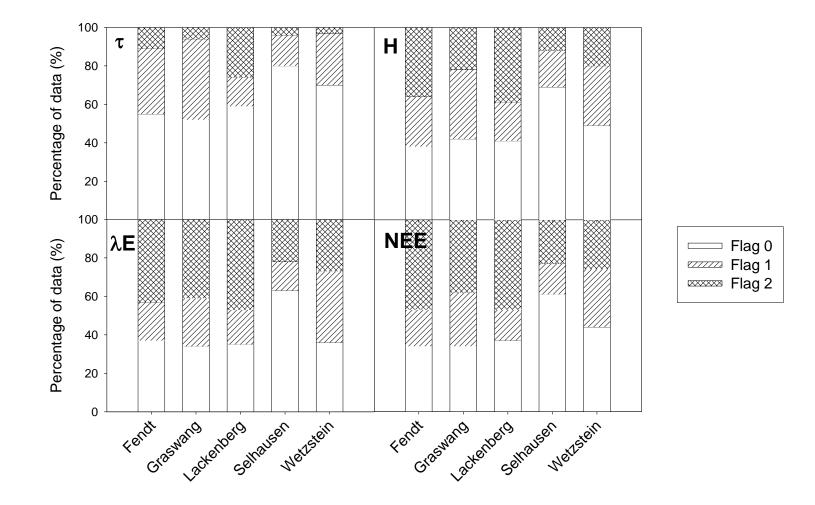
Overview of the test data sets

Site name	Operator	Ecosystem	Measurement height (a.g.l.*)	Sensor combination	Data period
Fendt	KIT	grassland in pre-	3.5 m	CSAT3/	25/07/2010 -
		alpine valley		LI-7500	23/08/2010
Graswang	KIT	grassland in pre-	3.5 m	CSAT3/	25/07/2010 -
		alpine valley		LI-7500	23/08/2010
Lackenberg	KIT	wind throw on	9.0 m	CSAT3/	25/07/2010 -
		low mountain range		LI-7500	23/08/2010
Selhausen	FZJ	agricultural land,	2.5 m	CSAT3/LI-	01/06/2011 -
		sugar beet		7500	30/06/2011
Wetzstein	MPI-BGC	Spruce forest on	30.0 m	Solent-R3/	15/07/2006 -
		low mountain		LI-6262	13/08/2006
		range			

* a.g.l.: above ground level

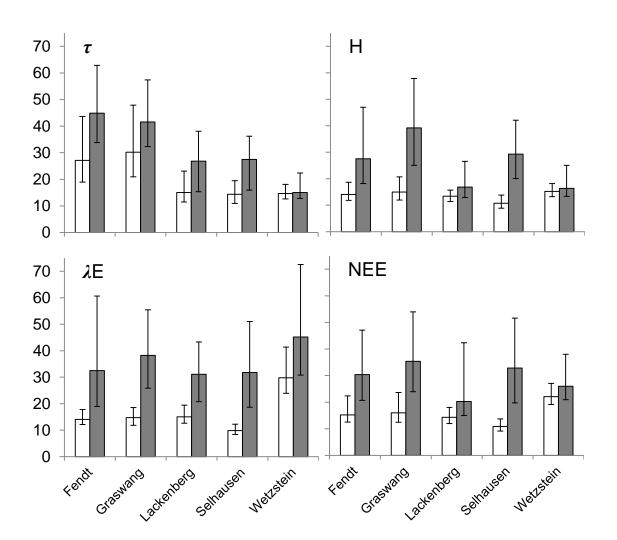
Results -Percentage of available data with flag 0, 1 and 2





Results – Relative random flux error (%) vs. quality flag







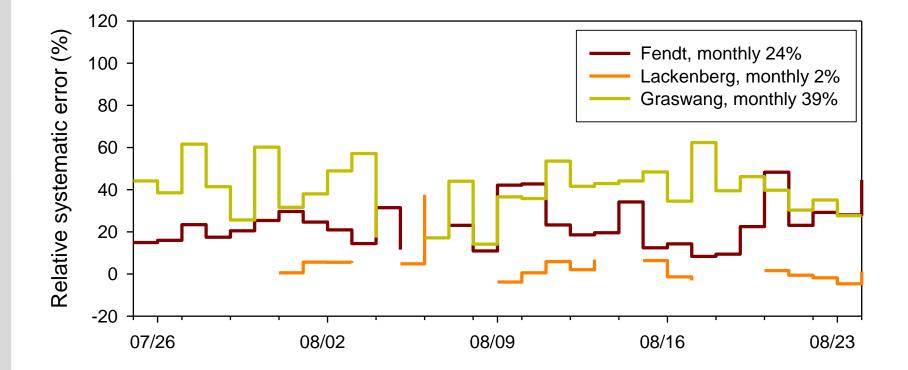
	Fendt	Graswang	Lackenberg	Selhausen	Wetzstein
	0.39	0.38	0.12	0.34	0.08
Н	0.54	0.64	0.31	0.95	0.30
λΕ	0.45	0.46	0.35	0.67	0.53
NEE	0.61	0.47	0.48	0.79	0.41

Even, if this error may be small in relation to the magnitude of the flux on average, this can be different for single estimates, e.g. the **maximum relative noise error** that we found for the five test data sets was **9%**.

Sturm P, Eugster W, Knohl A (2012) Eddy covariance measurements of CO₂ isotopologues with a quantum cascade laser absorption spectrometer. Agric Forest Meteor 152:73-82.

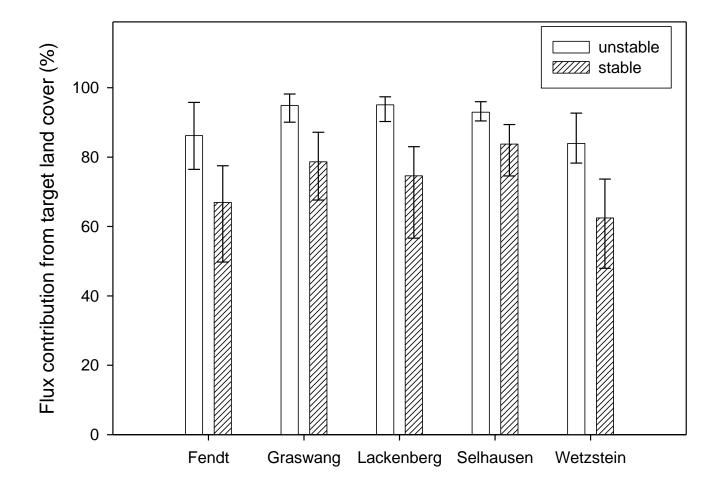
Results – Systematic errors in percent (%)





Results – Footprint: flux contribution from target

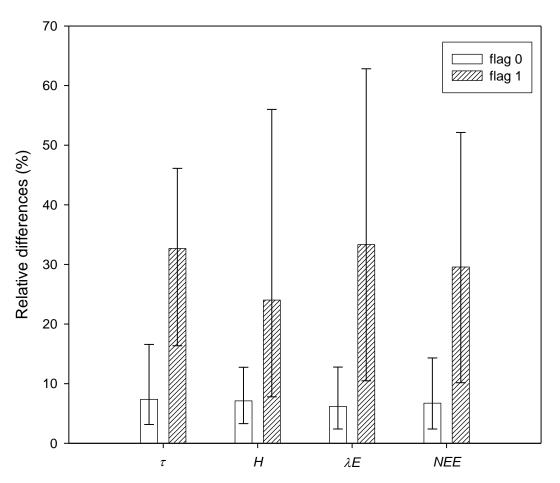




Possible further tests, not implemented in the quality assessment scheme



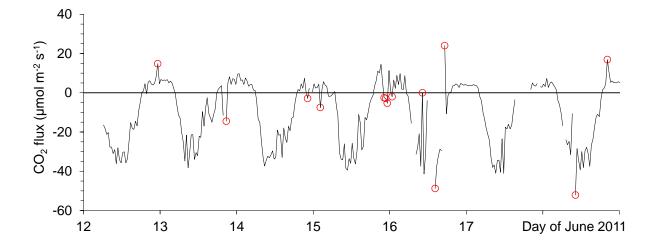
Stationarity-test based on detrending vs. block-averaging



Possible further tests, not implemented in the quality assessment scheme



Skewness and Kurtosis



Excerpt of CO₂ flux time series at the site Selhausen with visually identified strongly suspect fluxes (encircled).

Graf A, Schüttemeyer D, Geiß H, Knaps A, Möllmann-Coers M, Schween J, Kollet S, Neininger B, Herbst M, Vereecken H (2010) Boundedness of turbulent temperature probability distributions, and their relation to the vertical profile in the convective boundary layer. Bound -Layer Meteor 134:459-486

Vickers D, Mahrt L (1997) Quality control and flux sampling problems for tower and aircraft data. J Atmos Oceanic Technol 14:512-526.

Possible further tests, not implemented in the quality assessment scheme



Skewness and Kurtosis

Performance of the quality assessment scheme presented in this paper and several potential additional flagging schemes on visually determined suspect fluxes.

	presented	scheme	skewness and kurtosis flags			
	a) Flag 1 + 2	b) Flag 2	c) VM97	d) bimodal	e) K > β + 3	a) + d)
α: detections	3.0%	1.5%	1.7%	0.9%	0.3%	3.1%
β: missed	0.6%	2.1%	1.9%	2.7%	3.3%	0.5%
γ: false detection	24.4%	7.3%	30.8%	5.3%	8.2%	26.7%
δ: correct null	71.9%	89.1%	65.6%	91.1%	88.2%	69.6%
TSS: true skill score	0.59	0.33	0.16	0.20	0.01	0.59

Effectiveness of the quality assessment scheme



Results of the MAD-based outlier test (Papale et al. 2006) after application of the quality assessment scheme (number of detected values/available data after QC). Before QC, 1440 data were available per test data set.

	Fendt	Graswang	Lackenberg	Selhausen	Wetzstein
τ	1/1277	5/1348	0/1044	1/1383	2/1395
Н	1/916	7/1121	21/882	9/1262	19/1153
λΕ	2/820	5/850	7/762	13/1127	18/1059
NEE	3/757	9/888	8/765	7/1113	2/1064

Papale D, Reichstein M, Canfora E, Aubinet M, Bernhofer C, Longdoz B, Kutsch W, Rambal S, Valentini R, Vesala T, Yakir D (2006) Towards a standardized processing of Net Ecosystem Exchange measured with eddy covariance technique: algorithms and uncertainty estimation. Biogeosciences 3:571-583.

Conclusions



- An effective automatic quality assessment of long-term EC measurements was presented.
- This scheme assesses the data quality based on *fundamental* information contained in the measured data directly: no *a-posteriori* checks required.
- Regular/daily visual inspection of the data is still required.
- Expert knowledge can help to retain more data from the original data set.
- The presented quality assessment scheme will be the standard for ECmeasurements within TERENO. However, further development is possible when new scientific findings become evident.
- The algorithms are newly implemented in the software TK3.1 (freeware) and will be available as a pre- and postprocessor for the software package ECpack and as part of EddySoft.

Thank you for your attention!



