

Statistically based calibration/validation control of ATLID L1 data



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Abstract	Calibrating space-borne instruments	High Spectral Resolution Lidar (HSRL)			
We propose a set of parameters, which would characterize the behavior of the ATLID lidar system on a day-to-day basis using the L1 data as an input. With the help of this set we will trace:	• Calibration in the laboratory: high precision, repeatability, versatility. Not 100% consistent with the instrument after launch.	$I_m + I_p$ $\downarrow \qquad \qquad$			
(a) the stability of the detection chain for ATLID channels (Rayleigh, Mie, and the cross-polarized one);(b) the accuracy of cross-talk coefficients;	• Calibration in space using onboard sources and/or known external sources (stars, moon): typical for passive instruments.	$\gamma_{m} \stackrel{0.2}{\xrightarrow[]{0}{3}} \stackrel{1}{\xrightarrow[]{0}{3}} \stackrel{1}$			
(c) the stability of day- and nighttime noise; (d) the stability of the radiation detection for all atmospheric scenarios	• Calibration through collocation: ground-based stations, balloons,	$ \begin{array}{c c} \hline \\ \hline $			

and over the whole globe using a clustering algorithm applied to the scattering ratio (SR) histograms.

We define 11 parameters: 3 related to surface reflection, 6 related to stratospheric day- and nighttime noise for 3 channels, and 2 related to the SR histogram analysis. We demonstrate the feasibility of the approach using CALIOP L1 data for polarized and cross-polarized attenuated backscatter (ATB) components in 2008–2015.

Formulation of the problem

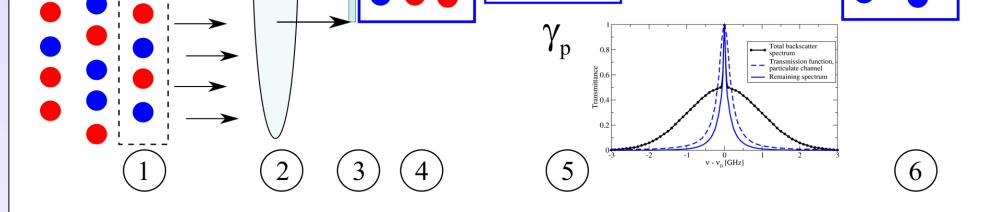
- Elements of spaceborne lidar, related to calibration:
- molecular channel
- aerosol channel

| <u>⊣</u> 5e-04 ⊢

- cross-polarized channel
- laser power measurement
- sending and receiving optics (alignment, coatings, degradation)
- data acquisition system (noise, electronic cross talk, etc)
- $L0 \rightarrow L1$ conversion requires knowledge of HSRL cross-talk coefficients (+ cross-talk for cross-polarized channel)
- How to detect drifts and offsets using only a flow of L1 data?
- Ideally, a set of parameters calculated on a day-to-day basis is needed: $\Delta_i < \text{threshold}_i \ (i=1...N)$

aircraft – compares the products (L2), involves $L0 \rightarrow L1 \rightarrow L2$ conversion, limited number of overlaps.

• Statistically based **quality control**: not equal to calibration, helps to identify issues in calibration and performance of the instrument, needs only a day-to-day flow of L1 data.



Schematic view of HSRL registering and treating the Rayleigh- (red) and Mie-scattered (blue) photons: (1) incoming flux, (2) telescope and optics, (3) HSRL splitting the components, (4) detectors, (5) transmission and spectra, (6) cross-talk free Rayleigh- and Mie-components

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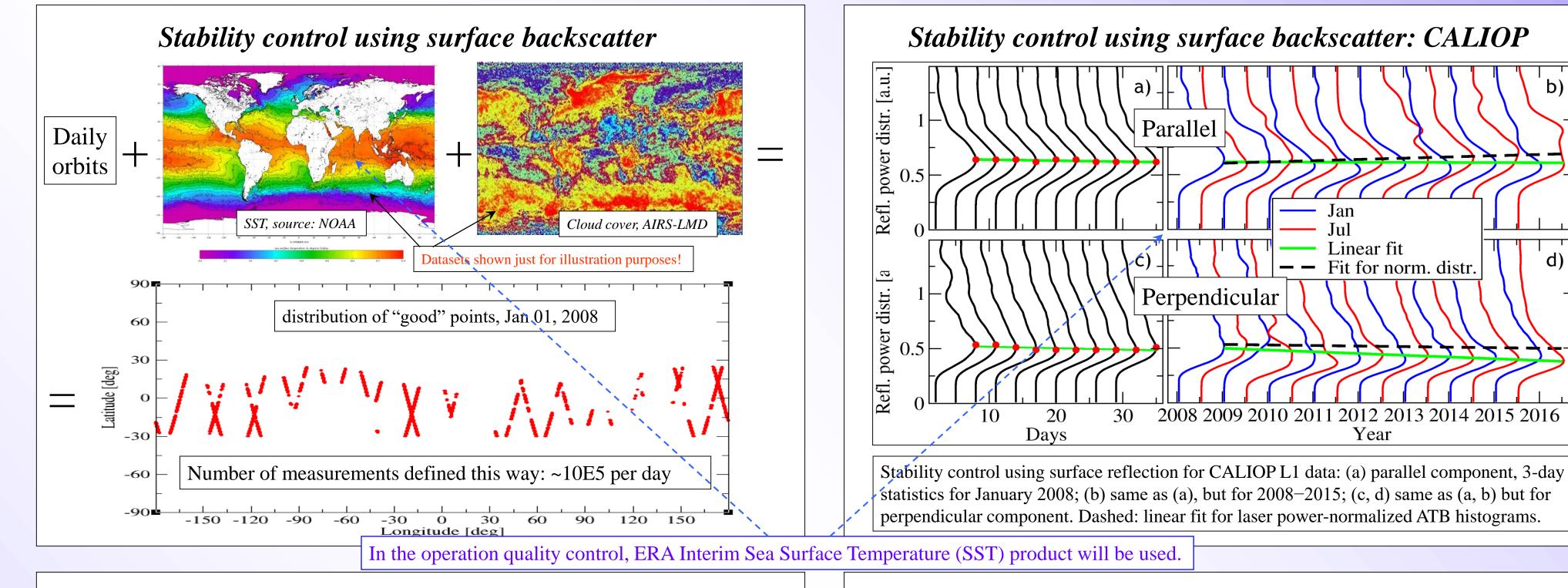
Linear fit

Year

Fit for norm. distr.

d)

ATTN: 6 Alert: 1



on= 109.5 lat= -15.5

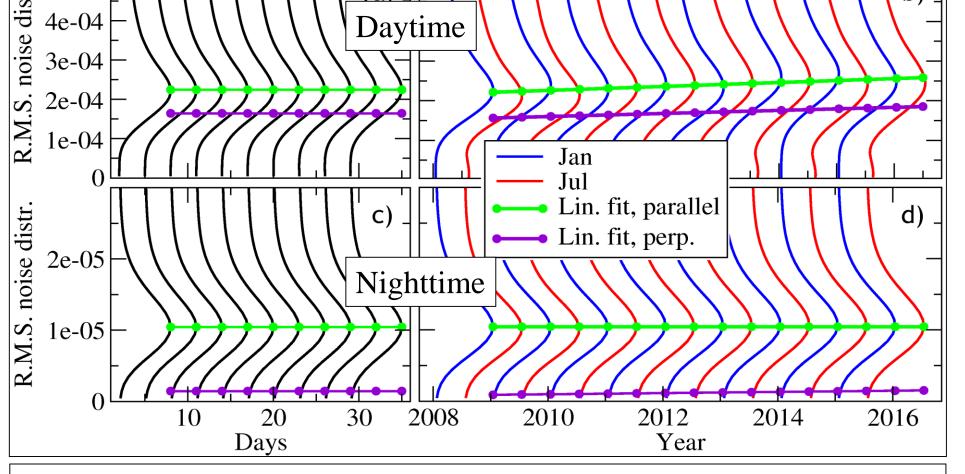
Scatterring ratio histograms: typical scenes

lon=-146.5 lat= 25.5

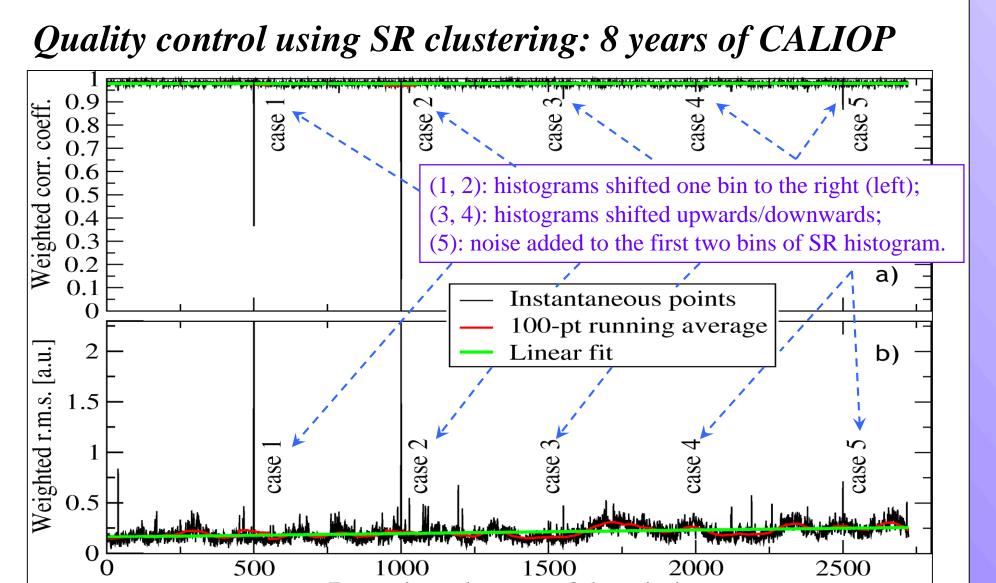
Example of geographical distribution of clusters

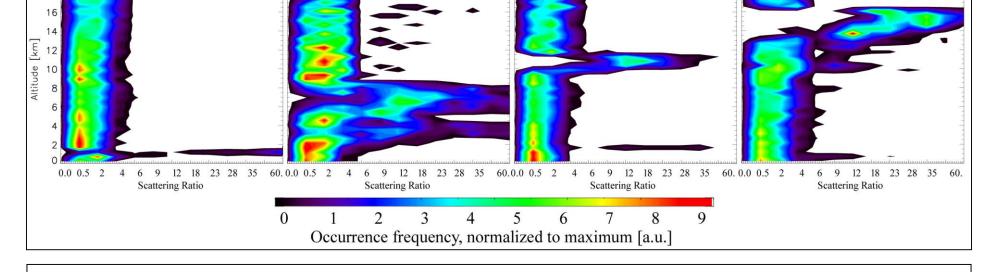
						-		
N_clusters=	6	•	1	1	7	20	26	42
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Stratospheric noise analysis: CALIOP



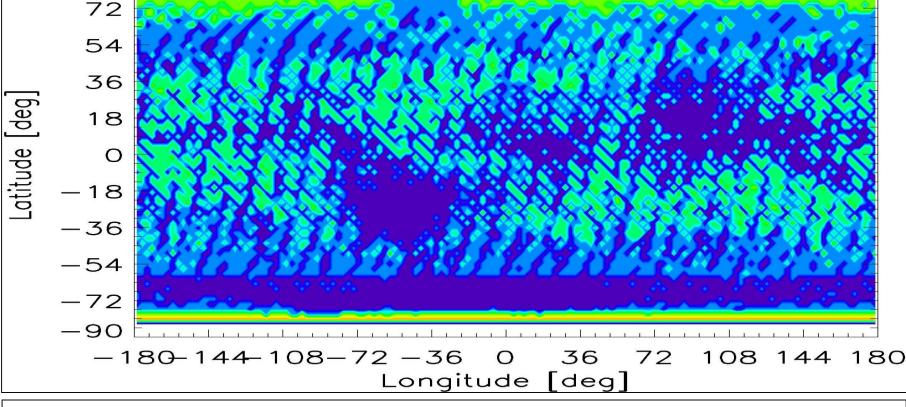
Histograms of stratospheric (35-40 km) signals for parallel component and linear fit of parallel (green line) and perpendicular (violet line) histogram maxima for: (a,b) daytime and (c,d) nighttime; (a, c) are zoomed versions of the first month of (b, d), respectively





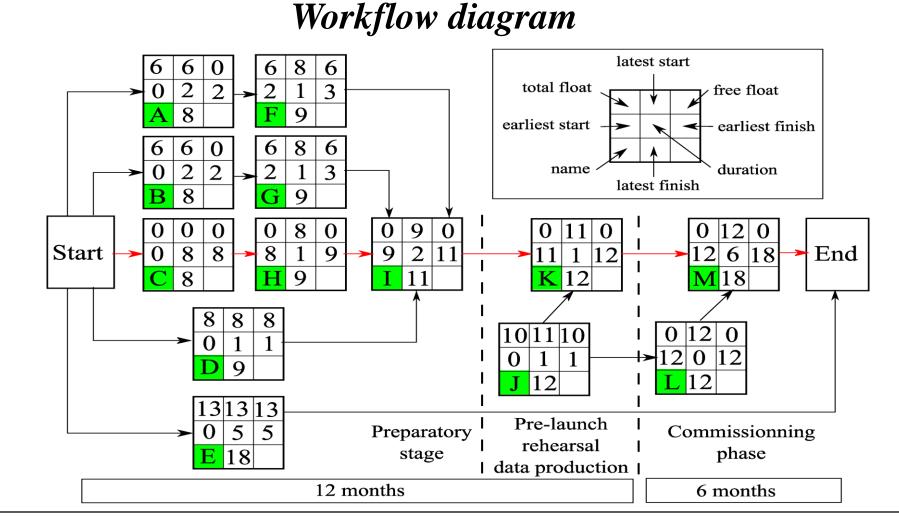
Examples of height-intensity SR histograms calculated for CALIOP: a) clear sky; b) multi-layer middle level cloud; c) high semi-transparent cloud; d) thick high cloud.

Approach already used by Marchand et al. [2008] for CloudSat and Chepfer et al., [2010] for CALIOP. Further development – **clustering** the same-type SR histograms: • "Clustering ... groups a set of objects in such a way that objects in the same group are more similar to each other than to those in other groups" [Everitt, 1993]. • Clustered histograms have good signal-to-noise ratio due to large statistics and they do not change in time rapidly. If new or rare phenomenon appears, it will be detected as such and it won't spoil the analysis.



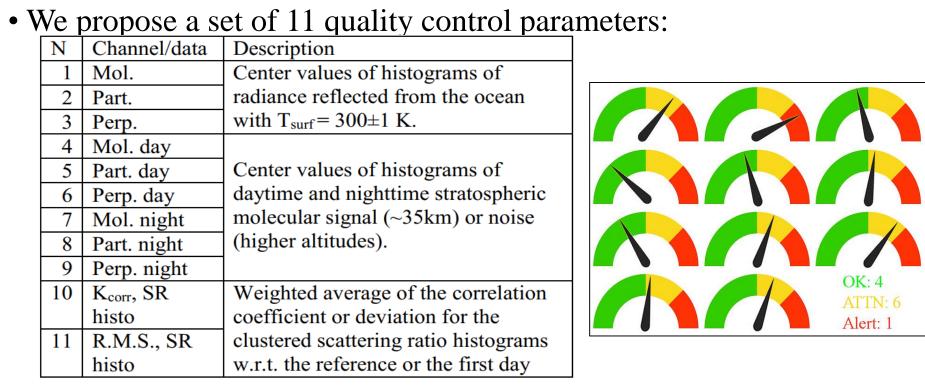
Results depend on the number of clusters prescribed to the algorithm. Optimal number is yet to be determined. Current estimate: 6-10.

Movement of air masses changes the location of clouds and aerosols, but the SR clusters remain the same (!). Natural day-to-day variability will determine the "allowed" quality control limits and can be estimated from CALIOP and simulations.



(A) surface reflection analysis \rightarrow 3 ATLID channels + ECMWF SST; (B) stratospheric analysis \rightarrow 3 ATLID channels; (C) SR histogram clustering algorithm \rightarrow ATLID; (D): Input interface \rightarrow

Summary



• We demonstrate the feasibility using 8 years of CALIOP data.

• The deliverables are:

(1) an operational quality control algorithm adapted to ATLID L1;



Daily quality criteria: (a) weighted correlation coefficient; (b) weighted r.m.s. of the difference. In 5 test cases the histograms were perturbed to simulate calibration issues.

ATLID data, output interface \rightarrow EarthCARE teams; (E): Documentation code transfer; (F, G, H)

Testing the (A, B, C) algorithms ; (I) Testing the joint package; (J) Pre-launch rehearsal; (K)

Testing the pre-launch data treatment; (L) Launch of the mission; (M) Operational analysis.

