Turbulent mixing in marine cumulus clouds experimental data analysis

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Atmospheric Physics Seminar 17.01.2025

Motivation

Clouds are very important objects present in the atmosphere. Their representation is regarded as the largest source of uncertainty in weather and climate models.

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This study is concerned with marine cumulus clouds. They are convective phenomena, and their size (both vertical and horizontal) is of the order of hundreds of meters.

Cloud processes

Cloud turbulence

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A useful concept for understanding the dynamics of a turbulent flow is the concept of an energy cascade.

Energy cascade

Figure: A schematic of an energy cascade

In this case ε is regarded as an energy flux going from the largest to the smallest scales. Several methods of obtaining ε from lower-resolution data exist, e.g.:

$$
E(\kappa) = C_K \varepsilon^{2/3} \kappa^{-5/3},
$$

\n
$$
D_{11}(r) = C_{11} \varepsilon^{2/3} r^{2/3}.
$$
\n(1)

Also, ε is connected to other quantities via Taylor's relation:

$$
\varepsilon = C_{\varepsilon} \frac{u'^{3/2}}{L},\tag{3}
$$

Deviations from HIT

Two deviations from the HIT theory are considered in this work:

- **1** Non-equilibrium turbulence
- **²** Anisotropic turbulence

Figure: A deviation from the -5/3 law

$$
E(\kappa) = C_K \varepsilon^{2/3} \kappa^{-5/3} \quad E(\kappa, t) = \overline{E}(\kappa, t) + \widetilde{E}(\kappa, t) \tag{4}
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\n
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\downarrow \text{ some derivation } \downarrow
$$

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$$
\sim \frac{\dot{\varepsilon}}{\varepsilon} \tag{5}
$$

↓ some more derivation ↓

e*k k*

$$
\frac{C_{\varepsilon}}{C_{\varepsilon}} \approx \left(1 + \frac{\widetilde{k}}{\overline{k}}\right) = \left(\frac{\text{Re}_{\lambda 0}}{\text{Re}_{\lambda}}\right)^{15/14} \to C_{\varepsilon} \approx \frac{1}{\text{Re}_{\lambda}}
$$
(6)

The derivation was done in Bos 2017.

The study of anisotropy includes the analysis of the ratio of magnitude of fluctuations in different directions e.g.:

$$
A_{u} = \sqrt{\frac{2\langle w'^{2} \rangle}{\langle u'^{2} \rangle + \langle v'^{2} \rangle}},
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as well as the ratio of TKE dissipation rates:

$$
A_{\varepsilon} = \frac{2\varepsilon_W}{\varepsilon_U + \varepsilon_V}.
$$
 (8)

Another way to study anisotropy is the analysis of the invariants of the *normalised anisotropic tensor*:

$$
a_{ij}=\frac{\overline{u'_i u'_j}}{2k}-\frac{\delta_{ij}}{3}.
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$$
I = \lambda_1 + \lambda_2 + \lambda_3 = 0
$$

$$
II = \lambda_1^2 + \lambda_1 \lambda_2 + \lambda_2^2
$$

$$
III = -\lambda_1 \lambda_2 (\lambda_1 + \lambda_2)
$$

Anisotropic turbulence

Figure: Turbulence triangle

A different approach

In this work, the clouds and cloud turbulence were also analysed based on a method that stems from the analysis of dynamical systems, namely Recurrence Plot and Recurrence Quantification Analysis.

Recurrence plot - definition

Let's define a vector \vec{x} which represents a state of a system.

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We can define a we define a recurrence matrix:

$$
R_{i,j} = \Theta(\epsilon - ||\vec{x}_i - \vec{x}_j||),
$$

where ϵ is called a recurrence threshold, \vec{x}_i is a vector at time *i*, and $\Theta(\cdot)$ is the Heavyside function, and $||\cdot||$ is a norm (in this case an L-2 norm).

Recurrence plot - example

Figure: Lorenz system

Recurrence plot - example

Figure: Recurrence Plot of the Lorenz system

Observation of RPs

1 Diagonal lines (parallel)

-> the evolution of states is similar at different times, the process could be deterministic

² White bands

-> nonstationarity, some states are rare or far from the normal, transitions may have occurred

³ Periodic patterns

-> cyclicities in the process

Observation of RPs

1 Vertical and horizontal lines/clusters -> some states do not change or change slowly for some time, indication for laminar states

- **²** Single isolated points -> heavy fluctuation in the process
- **³** Homogeneity:
	- -> the process is stationary
- **⁴** Fading to the upper left and lower right corners: -> nonstationarity; the process contains a trend or drift

RQA

Recurrence quantification analysis:

1 Recurrence Rate (RR)

$$
RR = \frac{\sum_{i,j=1}^{N} R_{i,j}}{N^2}
$$

² Determinism (DET)

$$
DET = \frac{\sum_{l=l_{min}}^{N} IP(l)}{\sum_{l=1}^{N} IP(l)}
$$

³ Laminarity (LAM)

$$
LAM = \frac{\sum_{v=v_{min}}^{N} vP(v)}{\sum_{v=1}^{N} vP(v)}
$$

RQA

Recurrence quantification analysis:

1 Average diagonal line length (L)

$$
L = \frac{\sum_{l=l_{min}}^{N} IP(l)}{\sum_{l=l_{min}}^{N} P(l)}
$$

² Trapping time (TT)

$$
TT = \frac{\sum_{v=v_{min}}^{N} VP(v)}{\sum_{v=v_{min}}^{N} P(v)}
$$

³ Ratio (RATIO)

$$
RATIO = \frac{DET}{RR}
$$

RQA - examples

Table: Values of selected quantities in the RQA for different systems

Data

The data used in this study come from the measurements done by the Twin-Otter aircraft during EUREC4A campaign.

Figure: A schematic showing the overview of the campaign

Data

Figure: A cloud

Data - problems

During the analysis, several problems with the data were identified.

Figure: An image showing example problems with the data

Data - conclusion

Table: Metadata. N_o – number of periods, ΣT_o – length of periods in minutes, *N^c* – number of clouds.

Data - example

Figure: An example time series showing three components of wind velocity, temperature, and liquid water content.

First, in order to calculate the fluctuating part of velocity (u') via Reynolds decomposition, we need to calculate the mean over a certain window, $τ_d$.

We also need to choose a window over which the statistics $(< \cdot >, \varepsilon$, etc.) are calculated, τ_a . Ideally, this window should be several times larger than τ_d .

Results - turbulence

Figure: A time series of basic turbulence quantities.

Results - turbulence

Figure: A time series of other turbulence quantities.

Results - nonequilibrium turbulence

Figure: A plot of C_{ε} vs Re_{λ}.

Results - anisotropic turbulence

Figure: Time series of *A^v* .

Time-dependent RQA

In this analysis, a time-dependent RQA was performed. That is, a matrix based only on a portion of the data was constructed, and then the standard RQA was performed. This way, time series of the aforementioned parameters were obtained.

In this case, a following vector was constructed:

$$
\vec{x} = [T, w', k]. \tag{10}
$$

Results - RQA

Figure: A time series of selected quantities in the RQA.

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Figure: An example time series with the LAM mask denoted by orange rectangles

Figure: An example time series of turbulence quantities with the LAM mask denoted by orange rectangles

Figure: Values of different quantities averaged with respect to 3 masks

Figure: Histograms of ε for different directions and masks

Conclusion

- **¹** An analysis of turbulence in and around cumulus clouds was performed.
- **²** Departure from HIT was observed. In my opinion, we need to look at cloud turbulence in terms of extended theories.
- **3** An RQA was done for the airborne time series.
- **⁴** A method to distinguish turbulent and non-turbulent segments was proposed. It has its advantages.

Outlook

This framework should be tested against a bigger dataset, but we are still waiting for the data.

A more in-depth analysis of parameters was performed for the PhD, but there are still possibilities to do more things with the RQA.

Literature

The literature and is available at BIP UW > DOKTORATY > Stanisław Król, or upon request.

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Thank you for your attention.