A new conceptual picture of the trade-wind transition layer* <u>Anna Lea Albright¹, Bjorn Stevens², Sandrine Bony¹, Raphaela Vogel³</u>

- Ecole Normale Supérieure, Ecole Polytechnique, Paris
- 2. Max Planck Institute for Meteorology, Hamburg, Germany
- 3. University of Hamburg, Germany

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1. Laboratoire de Météorologique Dynamique, CNRS, Sorbonne Université,

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Trade-wind atmosphere has characteristic vertical structure e.g., Malkus, 1958, Augstein, 1974, Yin & Albrecht, 2000



Studying vertical structure teaches us about physical processes producing this structure



Mark Rothko, Blue and Grey (1962)







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e.g., Malkus, 1958, Augstein, 1974, Yin & Albrecht, 2000

Common transition layer idealization (sharp gradients), In analogy with StCu regimes or dry convective layers

Specific humidity / gkg⁻¹

e.g., idealizations made in Lilly, 1968, Arakawa, Schubert, 1974, Betts, 1976, Albrecht, 1979, Stevens 2006

Sharp buoyancy gradients (green)

Direct numerical simulation results reproduced from Garcia & Mellado, 2014

How representative is this cloud-free structure?

altitude / m

Specific humidity / gkg⁻¹

Cloud-base cloud fraction measured from a lidar-radar synergy is small, 5.3±3.2% (Bony et al., 2022), so it appears reasonable that cloud-free transition layer structure could be the baseline

Most of the time, vertical gradients are smoother. How to define transition layer from profiles?

Apply height definitions, e.g., Canut et al., 2012, to observed thermodynamic profiles

Evidence for ~150-200 m thick transition layer between mixed and subcloud layer tops

- individually with transition layer

• Associate region between mixed layer and subcloud layer tops — that is better-mixed in θ_{y} than q, θ

Methodology for identifying transition layer given in Albright et al., Observed subcloud layer moisture and heat budgets in the trades, JAS, 2022, + implications for modeling subcloud layer thermodynamics

Transition layer thermodynamic gradients differ from those in mixed and cloud layers (810 dropsonde profiles composited by layer; mean depths)

What produces the observed transition layer structure thicker and with smoother gradients compared to jump structure?

Does life cycle (condensation-evaporation dipole) of very small clouds smooth vertical gradients in transition layer?

- About 60% of cloud bases (threehourly ceilometer data) and ~75% LCLs (from sondes) below transition layer top
- Another way of defining the transition layer is between cloud base and maximum cloud-base cloudiness level (cf. Vogel et al., 2022)

Cf. cloud based above the transition layer in Malkus, 1958; Augstein, 1974; but within transition layer in Neggers et al., 2009, Gentine et al., 2013

Test using denial of mechanism examine transition layer structure in large clear-sky areas

defined:

- 1. by eye, within patterns of cloud organization, identified from satellite images
- 2. as cloud-free over ~200 km of flight path (15 minutes of flying) using cloud flags and cloud top heights from WALES lidar
- 3. using large-eddy simulation output from Dauhut et al., 2022

Sharp gradients exist, but rarely, and in large clear-sky areas

Sharp gradients exist, but rarely, and in large clear-sky areas

Very shallow clouds are ubiquitous. Are they associated with smoother gradients?

Two cloud populations seen in satellite retrievals (e.g., Genkova et al. 2012, Leahy et al., 2012, Mieslinger et al., 2019), but with ~250-500 m observational uncertainties. Cf. also Vial et al, 2023

Proxy for transition layer structure

Select sharpest vertical gradient b/t 300-800 m

Specific humidity / gkg⁻¹

Large clear-sky areas (red) exhibit stronger vertical gradients

A similar picture in large-eddy simulation output

LES simulation output from Dauhut et al., 2022 QJRMS, 100 m (horizontal); 40 m (vertical)

A similar picture in large-eddy simulation output

LES simulation output from Dauhut et al., 2022 QJRMS,

100 m (horizontal); 40 m (vertical)

Find weak correlation b/t transition layer gradients & mesoscale subsidence (max. r~0.3 with ω_{2km})

Larger transition layer gradients with distance to cloud center of mass, evolving over time; confirmed by random forest analysis

Inferences from EUREC⁴A observations and LES

- Jump-like transition layer structure found in large (O(200 km)) cloud-free areas 1.
- 2. Strength of transition layer gradients only weakly associated with subsidence strength, maximizing at 2 km (r~0.3)
- Shallow population of clouds creates transition layer structure by a 3. condensation-evaporation dipole — active role for very shallow clouds that is missing from our previous conceptualization

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- Shallow population of clouds creates transition layer structure by a 3. condensation-evaporation dipole — active role for very shallow clouds that is missing from our previous conceptualization
- Inferences from mixed layer theory and mixing diagrams suggest that differences 4. in cloud-free and cloudy transition layer structures do not affect the rate of entrainment mixing, but rather the properties of the air incorporated into the mixed layer, primarily as a moistening (not shown)

form (cf. Neggers et al, 2015) and organize, and on what timescale?

Riehl et al, 1951

Interplay between very shallow and deeper shallow clouds, with each population growing its own layer (cf. Riehl, 1951, Stevens, 2007). Do smaller clouds make it easier for larger clouds to

Photo by Frédéric Batier, 2020

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- 2. Contribution to the energetics of entrainment mixing:

 - Cloud-free mixed layer theory, such as for entrainment rate closures, is still skillful with

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• Additional contribution to entrainment mixing based on ability to detrain condensate into the overlying stable layer, in addition to surface buoyancy fluxes, wind shear, radiative cooling appropriate modifications reflecting finite-thickness transition layer (cf. Albright et al., 2022)

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Stability conditions required for spontaneous growth of mesoscale moisture fluctuations and to power shallow circulations, e.g., $(\partial/\partial z)(\Gamma_{q_t}/\Gamma_{\theta_v}) > 0$, (cf. Eq. 26 in Janssens et al., 2023, and non-dimensional 'Chikira parameter', Eq. 41 of Bretherton and Blossey, 2017), appear to hold in

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<u>Related personal references</u>

Albright, A. L., Fildier, B., Touzé-Peiffer, L., Pincus, R., Vial, J., & Muller, C. (2021). Atmospheric radiative profiles during EUREC4A. Earth System Science Data, 13(2), 617-630 Albright, A. L., Bony, S., Stevens, B., & Vogel, R. (2022). Observed subcloud layer moisture and heat budgets in the trades. Journal of the Atmospheric Sciences Albright, A. L., Stevens, B., Bony, S., & Vogel, R (2023). A new conceptual picture of the trade-wind transition layer. Journal of the Atmospheric Sciences

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Additional slides

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(3) No longer cloudy, but influenced by dissipated clouds

(2) Cloudy convective boundary layer

Jumps in Thibaut's LES

Do differences in transition layer structure matter for mixed layer state & surface fluxes?

- Inferences from mixed layer theory and mixing diagrams (not shown, following Paluch, 1979) suggest that the observed transition layer structure does not strongly affect the rate of entrainment mixing
- Rather, it influences the properties of the air incorporated into the mixed layer, primarily as a moistening

Contributions to energetics of entrainment mixing

Convert turbulence energy to potential energy

e.g., Ball, 1960, Lilly, 1968, Betts, 1973, Tennekes, 1973, Deardorff, 1974, Stull, 1976, Stevens 2006

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A_e: entrainment efficiency of surface turbulence source (constant)

 $A_e = 0.2? 0.4?$

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$A_e = 0.43^*$

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Cloud liquid water flux contribution to $A_e \sim 0.4$

A short side project

Can we predict transition layer gradients based upon environmental variables?

- **Random forest** or XGBoost (machine learning) algorithms

<u>100 m (horizontal); 40 m (vertical) model output</u>

• 9 variables considered: {q, θ , wind speed in mixed layer; vertical velocity at different altitudes; integrated cloud liquid water content; distance to cloud 'center of mass', cloud base height, cloud top height}

Preliminary take-away:

Algorithm has some predictive skill for maximum transition layer vertical gradient

Most important environmental variables: but, a cold pool imprint rather than a predictive feature?

Features

Visualizing Important Features

Cold pool signature? Regions $\theta \leq 297.5$ in contours

Skill increases when re-running analysis outside cloud region, and distance to cloud center of mass is most important variable

Q: Can we use machine learning techniques to identify important features associated with transition layer gradients?

- LES simulation output: 100 m (horizontal), 40 m (vertical) Dauhut et al., 2022 QJRMS
- Target: transition layer gradient, quantified as maximum first difference (over 40 m) b/t 400-1000 m
- 9 features considered: q, θ, wind speed at 500 m; vertical velocity at 500 m or 2km; integrated cloud liquid water content; distance to cloud 'center of mass', cloud base height, cloud top height
- Random forest or XGBoost (machine learning) algorithms used because they are more interpretable than deep learning, and allow for nonlinear relationships unlike multiple linear regression

Distance (km)

Initial approach has predictive skill, but feature importance reflects cold pool signature, rather than meaningful predictor

Distance to cloud emerges as most important feature when considering areas outside cold pools*, and skill improves

those influenced by precipitation and cold pools

