Can we link theory to observations in natural flows?

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What When Batchelor in Chemistry (Uni Trieste, IT) 2001-2005 Internship at INOGS (IT) 2005 Master in Physics (Uni Trieste, IT) 2005-2008 Internship at Elettra Synchrothron (IT) 2008 PhD at KNMI/Uni Utrecht (NL) 2009-2013 Post-doc at NIOZ (NL) 2013-2016 Post-doc at EPFL (CH) 2016-2018

- 17 peer-reviewed papers and proceedings
- Lots of (open-source) software
- Several datasets collected
- Conferences, workshops (EGU, Euromech,...)
- Summer schools: DAMTP Cambridge, Alpine Summer School

• Can we link theory to observations?

An overview of my work at NIOZ and EPFL

Context: Increasing amount and quality of observational data.

Opportunity to obtain a "statistical" description.

- Focus on full dataset instead of single events.
- In practice: spectra, structure functions, PDFs, PCA,...

Statistics of temperature, velocity,...

- scale-dependence of variability;
- intermittency (e.g., time/space dispersion of turbulence events);
- hints on underlying physical mechanisms;
- identification of different regimes at different scales.

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- relatively well understood theory (single-process level);
- laboratory studies.

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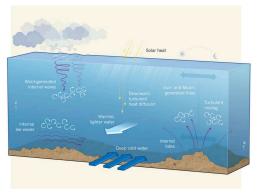
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- laboratory studies.

In the field:

- We cannot control what we observe in the field,
 - e.g. control parameters are variable / undefined.
- Statistics can help extracting information from "noisy" data.

Turbulent transport in the deep ocean (NIOZ)

- Vertical transport in the ocean interior is poorly understood
- Hypothesis of mixing "hot-spots" with sloping bottom
- Sparse observations, poorly understood dynamics

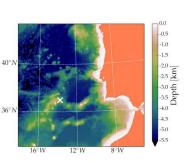


MacKinnon, 2013

Data

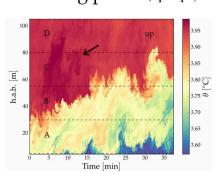
Latitude
Longitude
Max. depth
Min. height above seafloor
Seafloor slope
Number of sensors
Vertical spacing
Depth range
Deployment
Recovery
Sampling rate

36° 58.885′ N 13° 45.523′ W 2205 m 5 m 9.4° 144 0.7 m 100.1 m 13 Apr 2013 12 Ago 2013 1 Hz

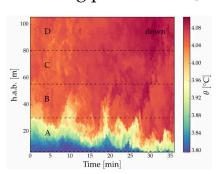


Data

Cooling phase (upslope)



Warming phase (downslope)



Generalised structure functions (GSF)

GSFs provide a way to characterise variability in a flow:

$$\gamma_q \equiv \gamma_q(r) = \langle |\Delta_r \theta|^q \rangle$$

So-called "scaling ranges" have been predicted by theory and observed in the laboratory:

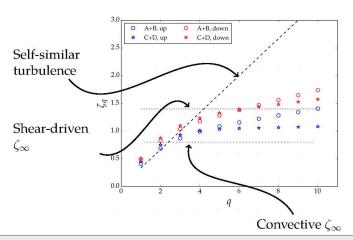
$$\gamma_q \sim r^{\zeta(q)}$$
.

- $\zeta(q) = q/3$ if turbulence were fully self-similar (non-intermittent), for r within the "turbulence inertial range".
- In reality, $\zeta(q) = \zeta_{\infty}$ for q > 10 (saturation):
 - Grid turbulence, shear driven $\rightarrow \zeta_{\infty} \approx 1.4$
 - Convective turbulence, buoyancy driven $\rightarrow \zeta_{\infty} \approx 0.8$

Generalised structure functions (GSF)

...many steps afterwards...

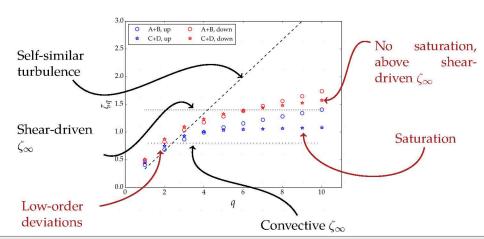
Scaling exponent within the turbulence inertial range.



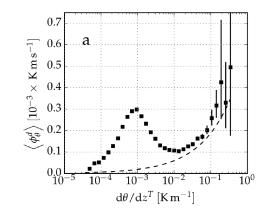
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Scaling exponent within the turbulence inertial range.



Flux-gradient relation



Long-term averaging enables to identify simple *mean* behaviour in an otherwise highly variable environment.

Conclusions I

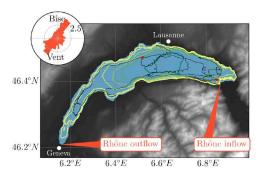
Evidence of convection in field observations

- Previously hypothesised, but hard to measure.
- Implications for efficiency of turbulent transport:
 - Implications for transport of heat, CO₂, nutrients,...
- Generalised structure functions enable to identify points of contact between laboratory and field results...
- ...and discrepancies!
- Much more not shown here.
- A. A. Cimatoribus and H. van Haren. Temperature statistics above a deep-ocean sloping boundary. J. Fluid Mech., 775:415

 –435,
 2015.
- A. A. Cimatoribus and H. van Haren. Estimates of the temperature flux-temperature gradient relation above a sea floor. J. Fluid Mech., 793:504–523, 2016.

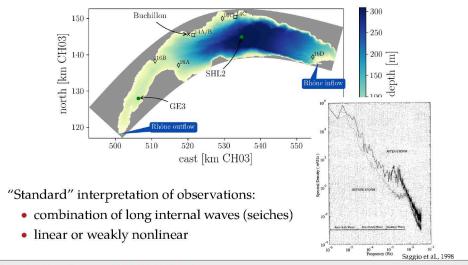
Transport in coastal areas (EPFL)

- In a linear, rotating flow, cross-isobath velocity is zero.
- How does cross-shore transport take place?



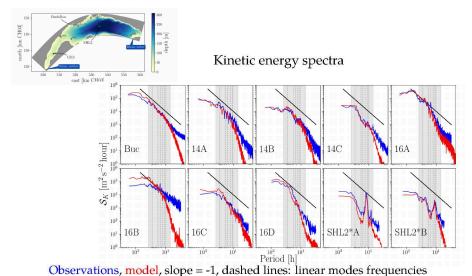
Velocity spectra

Lake Geneva



Velocity spectra

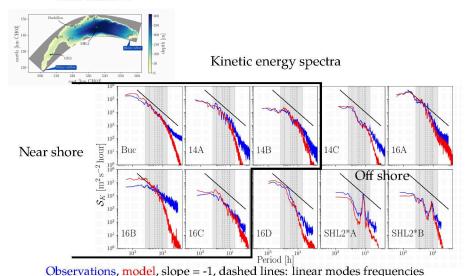
Lake Geneva



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Velocity spectra

Lake Geneva



Conclusions II

Nonlinearity cannot be neglected.

- Linear models predict sharp spectra, which are only observed off-shore,
 - or by considering short time intervals
- Broad spectra from longer time series suggest strongly nonlinear dynamics (more "turbulence"-like)
 - Confirmed by numerical modelling results.
- Classical tools like PCA, struggle to capture relevant dynamics in the highly variable, strongly forced, Lake Geneva.
- Andrea A. Cimatoribus, U. Lemmin, D. Bouffard, and D. A. Barry. Nonlinear Dynamics of the Nearshore Boundary Layer of a Large Lake (Lake Geneva). J. Geophys. Res. Oceans, in press, 2018.

Conclusions

- Field observations are growing in size and quality.
- A statistical description allows testing theories in a natural (uncontrolled) environment,
 - whose overall, mean behaviour is usually the most interesting one.
- Sometimes, statistical quantities can surprise:
 - Simple behaviour out of highly turbulent environments
 - Nonlinear behaviour (instabilities? vortices?) in a low energy environment
 - from very common power spectra!

Thanks for listening.



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[La Palma, Islas Canarias]