

1. Introduction

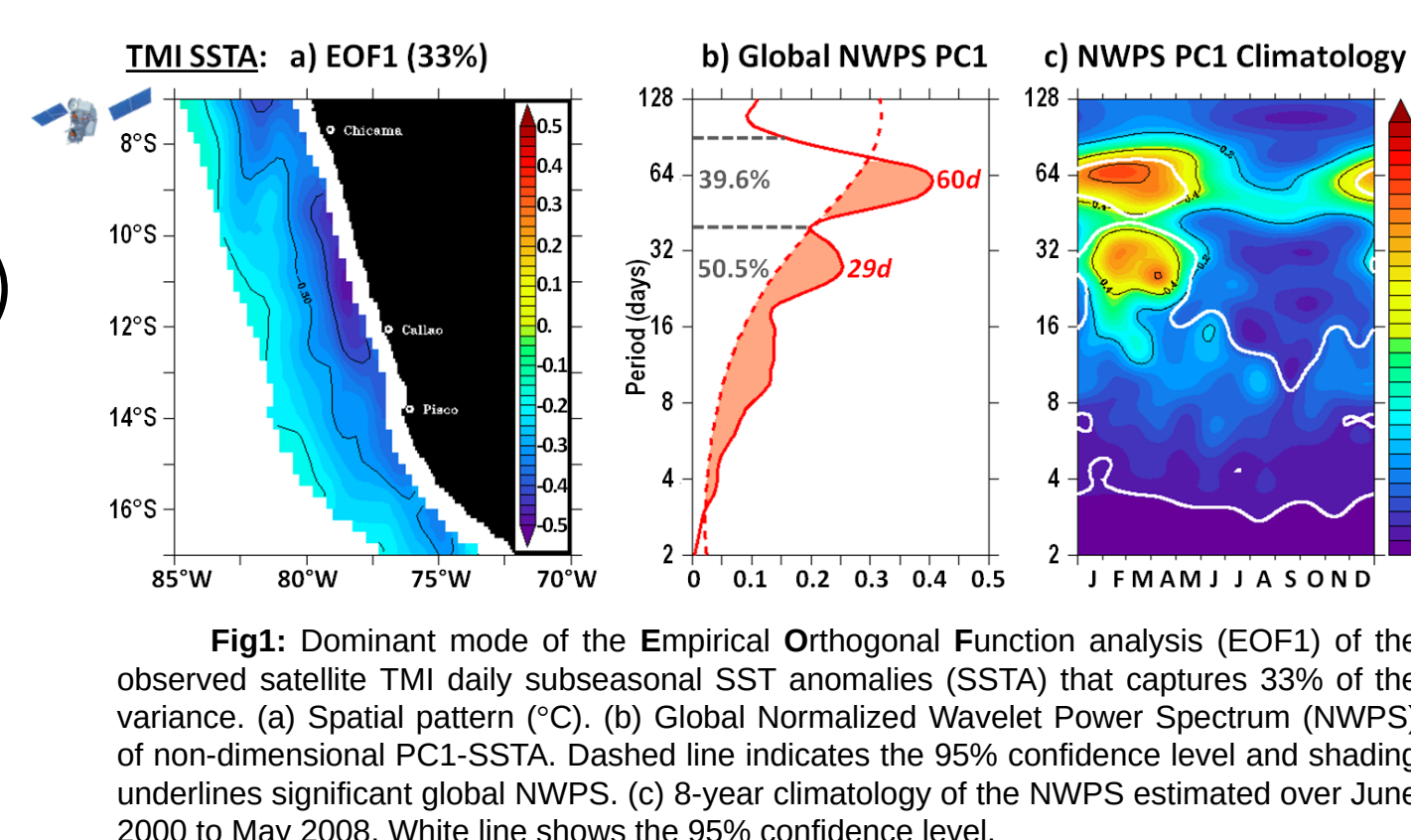
The Sea Surface Temperature (SST) **intraseasonal** variability ([40-90] days) along the coasts of Peru (**Fig1**) is commonly attributed to the efficient oceanic connection with the equatorial variability.

➤ In agreement with *Dewitte et al (2011)*, the **subseasonal SST** variability off Peru (**Fig1a**) consists in a narrow coastal fringe of SSTA with amplitudes decreasing from the coast to the offshore ocean. It highlight **two regimes of variability (Fig1b)**:

➔ A wind forced sub-monthly regime (<40 day⁻¹) that explains 50%

➔ An **intraseasonal regime** (40-90 day⁻¹)

➤ Here, we focus on the **intraseasonal SST regime** (= 40% of subseasonal var). It shows a significant peak at ~60 days with a marked seasonal dependence, prominent in Austral Summer (**Fig1c**)



5. Local forcing and processes

Fig9: Time-lagged composites of the intraseasonal anomalies of ROMS^{WQ} simulation averaged between 16°S-8°S as a function of the distance from the coast (in 10° km) for intraseasonal cold events: temperature in the mixed layer (T_{MLD} in °C), along shore wind stress (100dyn/cm²), surface net heat fluxes (Q_{net} in W/m²) and mixed layer depth (h in m).

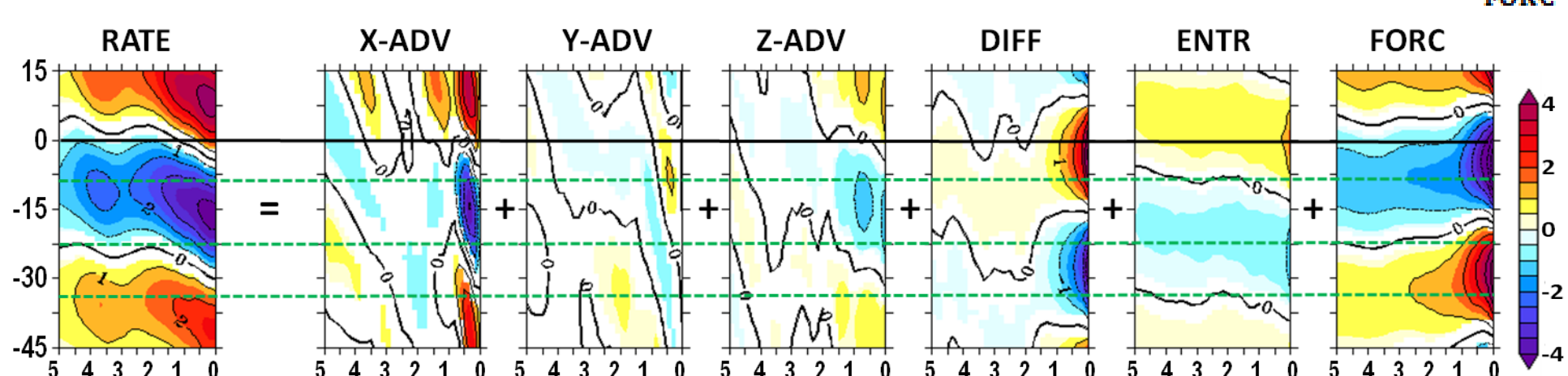
To filter out the internal model variability and provide an overview of the time sequence of intraseasonal cold events:

- We used a composite analysis of strongest intraseasonal SST events.
- Of latitudinal averages (16°S-8°S)
- Results are displayed as a function of the distance to the coast (**Fig9-10**).

- Black horizontal lines (—) ⇔ Mature phase of the event (@lag=0day), when the rate of change of T_{MLD} is zero.
- Green dashed lines (---) ⇔ characteristic of the wind stress activity lag=8day ⇔ Max. of the upwelling favorable wind (>0.8 dyn.cm⁻²)

2) lag=-22day ⇔ Min in alongshore wind stress @ lag=-22day)
lag=-33day ⇔ Max of downwelling favorable wind stress

$$\frac{\partial T}{\partial t} = -(\mathbf{u} \cdot \nabla_x T) - (\mathbf{v} \cdot \nabla_y T) - \frac{(\mathbf{w} \cdot \nabla_z T)}{h} - \frac{(\mathbf{w} \cdot \nabla_z T)}{h} \times ((T) - T_{ref}) + \frac{p C_p h}{h} \frac{\partial T}{\partial t}$$



The analysis of the mixed-layer processes (**Fig10**) highlights the active role of the advection processes associated with **Ekman dynamics**. Still, due to the shallow mixed-layer during this season, **diabatic processes** have a significant contribution (~30%) which is influential on the evolution of SST anomalies. In particular, they favor a faster response of SST to wind stress, reducing the quadrature between SST and wind stress anomalies expected from a conceptual advective model ($\frac{dT}{dt} = -\mathbf{w} \cdot \frac{\partial T}{\partial z}$).

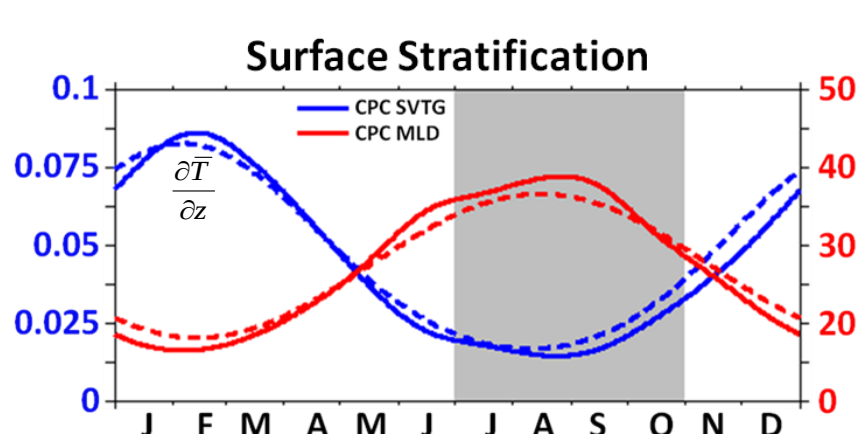


Fig11: Coastal (0.5° coastal band averaged) Surface Vertical Thermal Gradient (SVTG; difference between SST and temperature at 50m depth, blue, left scale, °C/meter) and Mixed Layer Depth (MLD, red, right scale, meters) averaged between 16°S-8°S for ROMS^{CR} (plain lines) and CARSS²⁰⁰⁹ (dashed lines). Grey shading is the July to October season.

In austral winter, despite the strong wind stress variability (**Fig2**), the vertical advection is reduced: ➔ **Surface stratification** characteristics = magnitude of the Surface Vertical Temperature Gradient (SVTG, **Fig11**) determines the efficiency by which wind stress anomalies force SST anomalies through advection processes (*Goubanova et al. (2013)*).

2. Objectives

What are the **forcing mechanisms** of the **intraseasonal SST variability** off central Peru (8°S-16°S) during the 2000-2008 period?

➔ **Locally forced**

Wind stress, Heat Fluxes?

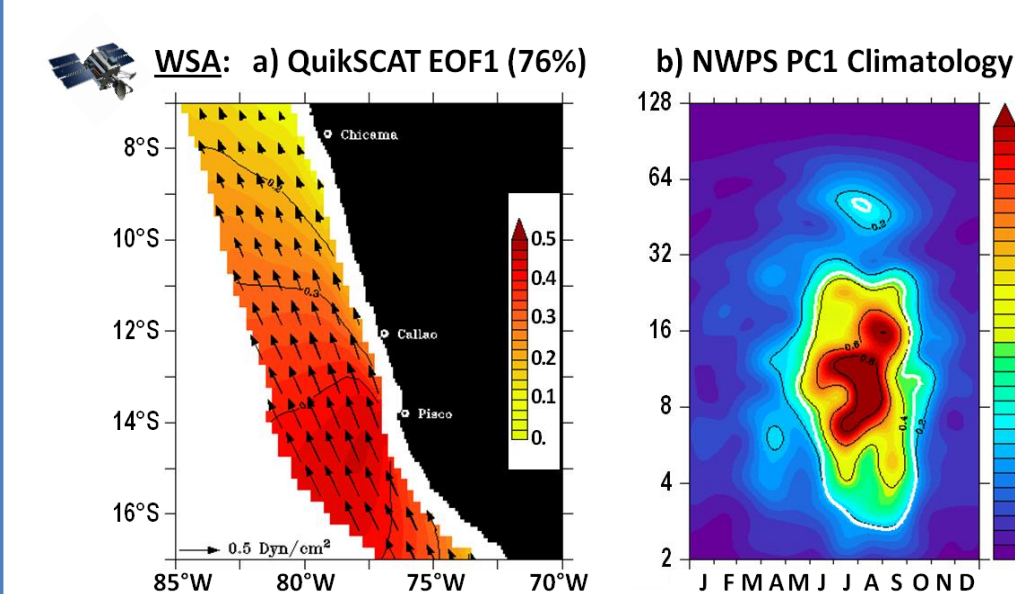


Fig2: First mode of the bivariate EOF analysis of the subseasonal anomalies of the surface wind stress components from QuikSCAT data: (a) Spatial pattern (dyn/cm²). (b) 8-year climatology of the NWPS for the wind stress PC1 (PC1-WSA) estimated over June 2000 to May 2008. White line shows the 95% confidence level.

➤ **Local wind stress** forcing shows upwelling favorable pattenr that is prominent in Austral Winter and barely significant at intraseasonal timescales **Fig2**

➤ Analysis of **Intraseasonal Equatorial Kelvin Wave (IEKW)** activity at 100°W shows that the remote equatorial forcing shares the temporal characteristics of the intraseasonal SST variability off Peru **Fig3**

➔ **Remote forcing**

Connection with the equator

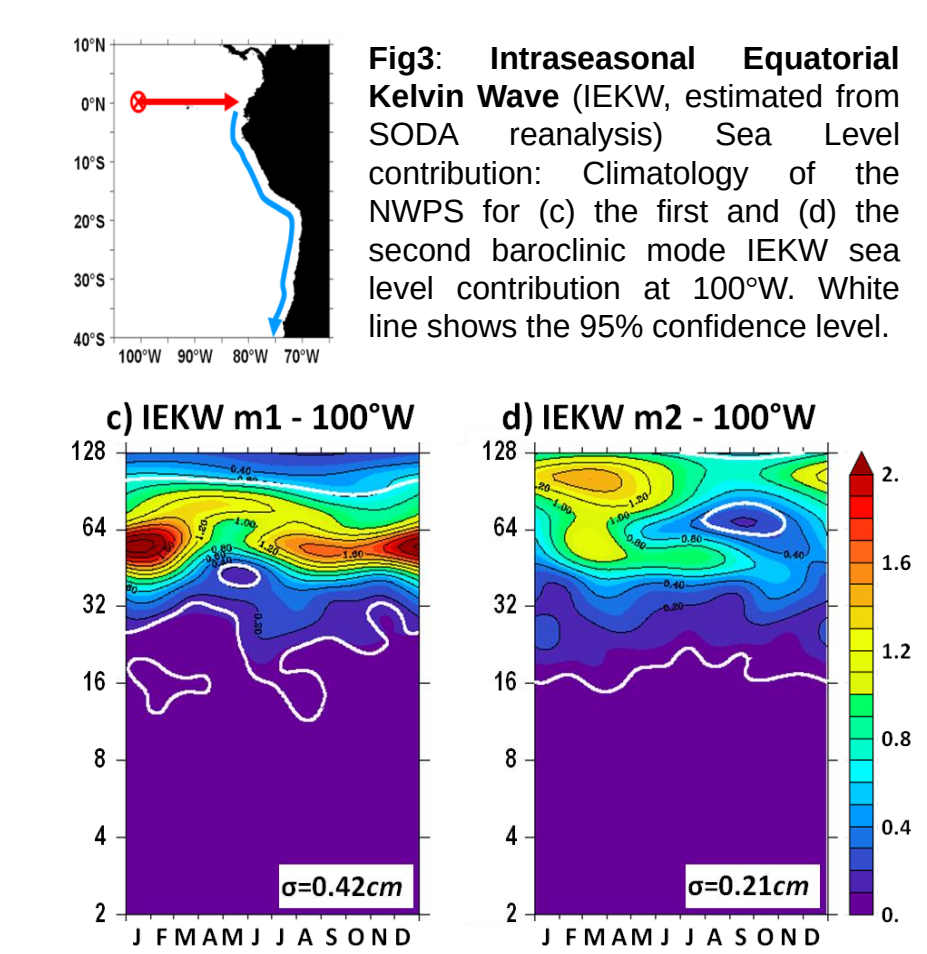


Fig3: Intraseasonal Equatorial Kelvin Wave (IEKW, estimated from SODA reanalysis) Sea Level contribution: Climatology of the NWPS for (c) the first and (d) the second baroclinic mode IEKW sea level contribution at 100°W. White line shows the 95% confidence level.

4. Remote equatorial forcing quantification

➔ Comparison of the model experiments reveals that the **intraseasonal SLA variability** off central Peru is accounted for by almost the sole equatorial forcing **Fig7**

➔ The ratio of variance between ROMS^E (ROMS^{WQ}) and ROMS^{CR} estimated for the [40-90] day⁻¹ scales for central Peru coastal SLA (averaged between 16°S-8°S within the 0.5° coastal fringe) is 95% (24%).

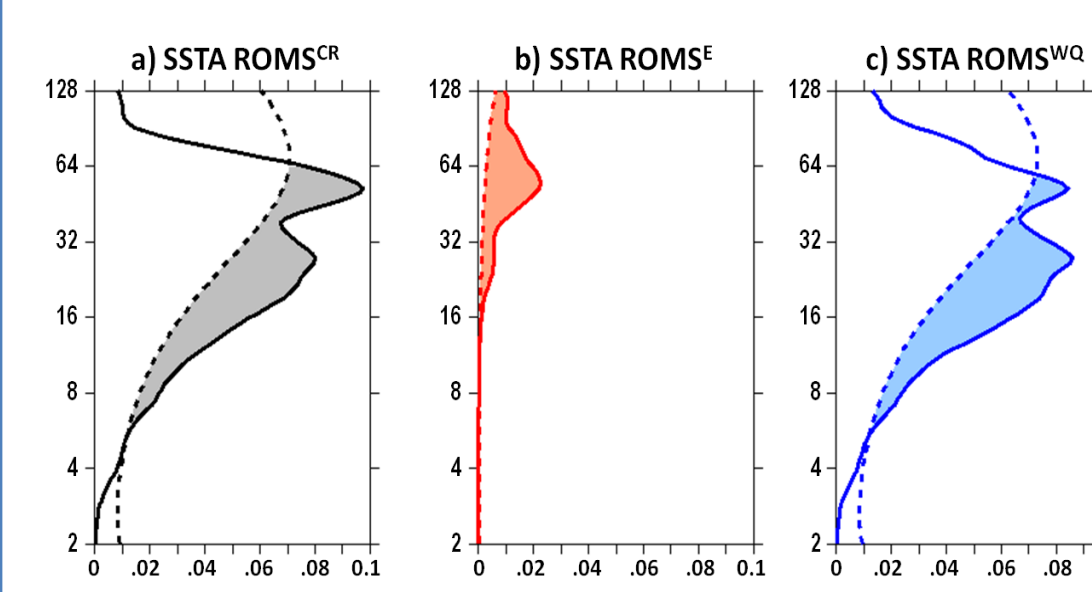


Fig7: Coastal (0.5° coastal band averaged) SSH averaged between 16°S and 8°S: (top) Global NWPS of subseasonal SSH for (a) ROMS^{CR} simulation, (b) ROMS^E simulation and (c) ROMS^{WQ} simulation. Unit is cm². Dashed lines indicates the 95% confidence level.

➔ Connection with the remote equatorial variability is only marginally influential on the **intraseasonal SSTA variability**, as it represents only ~23% of the intraseasonal SST variability off Central Peru in 2000-2008 (**Fig3**) !
➔ Conversely, the local forcing explains most of the intraseasonal SST variability along the coast of central Peru, since the spectrum for ROMS^{CR} and ROMS^{WQ} are comparable (ratio=93%) for the [40-90] day⁻¹ scales.

Since the seasonal dependence of the intraseasonal wind stress activity (**Fig1c**) is different from the one of the SST (**Fig2b**), the model results are counter-intuitive. This calls for investigating the mixed-layer processes associated with the local atmospheric forcing.

6. Summary and conclusions

Despite evidences of **clear propagations** of coastal trapped waves of equatorial origin observed in the thermocline fluctuations until 30°S (**Fig12**) and the comparable marked seasonal cycle in the intraseasonal Kelvin wave and coastal SST variability (*i.e.* peak in Austral summer) off central Peru, the **remote equatorial forcing only accounts for ~20% of the intraseasonal SST regime**, which instead is mainly forced by the local winds and heat-fluxes.

A heat budget analysis further reveals that during the Austral summer, despite the weak along-shore upwelling (downwelling) favourable wind stress anomalies, significant cool (warm) SST anomalies along the coast are to a large extent driven by **Ekman-induced advection**. This is shown to be due to the shallow mixed layer that increases the efficiency by which wind stress anomalies relates to SST through advection.

Thus, the **SST intraseasonal regime** during the Austral summer is **mostly wind-driven** despite the very weak wind stress variability (*i.e.* ~6 times weaker than in Austral winter).

Diabatic processes also contribute to the SST intraseasonal regime, which tends to shorten the lag between peak SST and wind stress anomalies compared to what is predicted from a simple advective mixed-layer model.

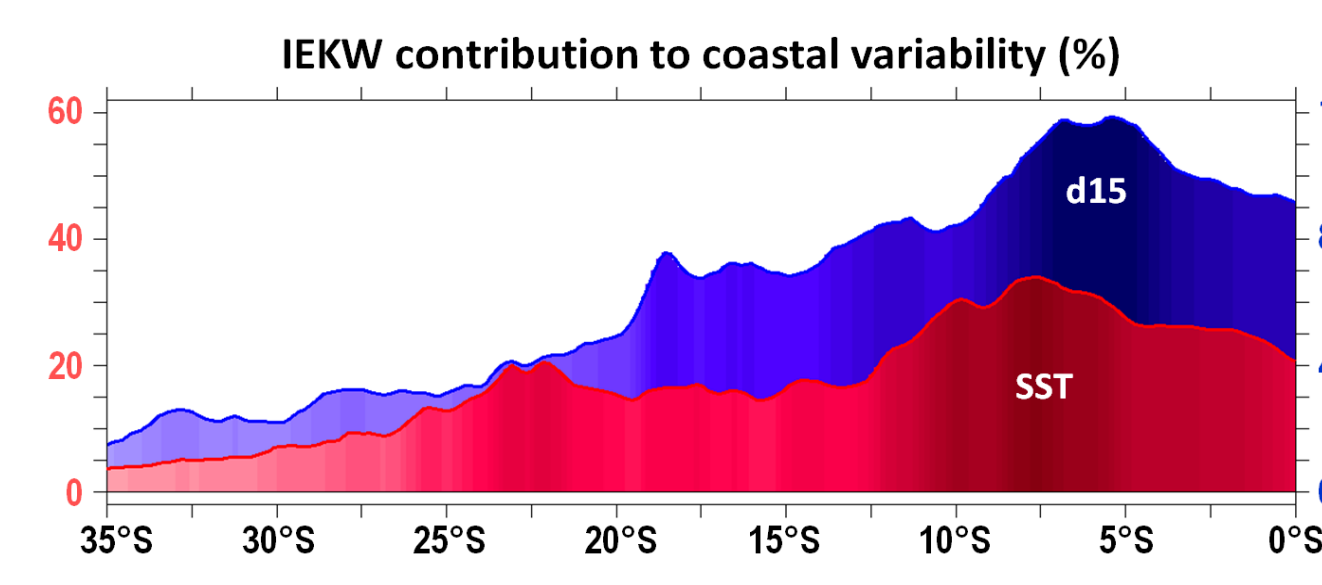


Fig12: Intraseasonal variability ratio (%) of variance between ROMS^E simulation and ROMS^{CR} as a function of latitude for the SST (red, left scale) and the depth of the 15°C isotherm (d15, blue, right scale) averaged in a 0.5° coastal band.

3. Methodology

Our approach is based on the experimentation with the regional ocean model ROMS

Model Configuration :

- Domain: [40°S-15°N ; 100°W-70°W]
- Horizontal resolution : 1/6°
- 32 (sigma) vertical levels
- Bathy GEBCO_08

Model Forcing :

- Wind stress : **QuikSCAT daily (CERSAT)**
- Heat/Water Fluxes : **Bulk formula with daily ERA-I (CERSAT)**
- Boundary Condition : **SODA (5 days)**
- Spin-up : 3 years (2000)
- Simulation over 2000-2008 with daily outputs

Model Validation :

Our confidence in the model configuration results in its realism as evidenced by the detailed validation of the model control run experiment (ROMS^{CR}) which indicates that it is **skillful** in simulating most aspects of the **mean state** and the **intraseasonal variability**.

• **Fig4 :** The model has a good mean SST, with averaged bias < 0.5°C.

• Compared to **Fig1**, **Fig5** shows that the model captures the main characteristics of the dominant mode of intraseasonal SST variability off central Peru, with correlation (RMSdiff) between model and observed PC1-SSTA (**Fig5d**) of 0.88 (0.25°C).

• **Fig6 :** The agreement between model Sea Level and AVISO data is statistically significant along the equatorial wave guide and its coastal extension, highlighting the efficient oceanic connection with the equatorial variability at intraseasonal times scales.

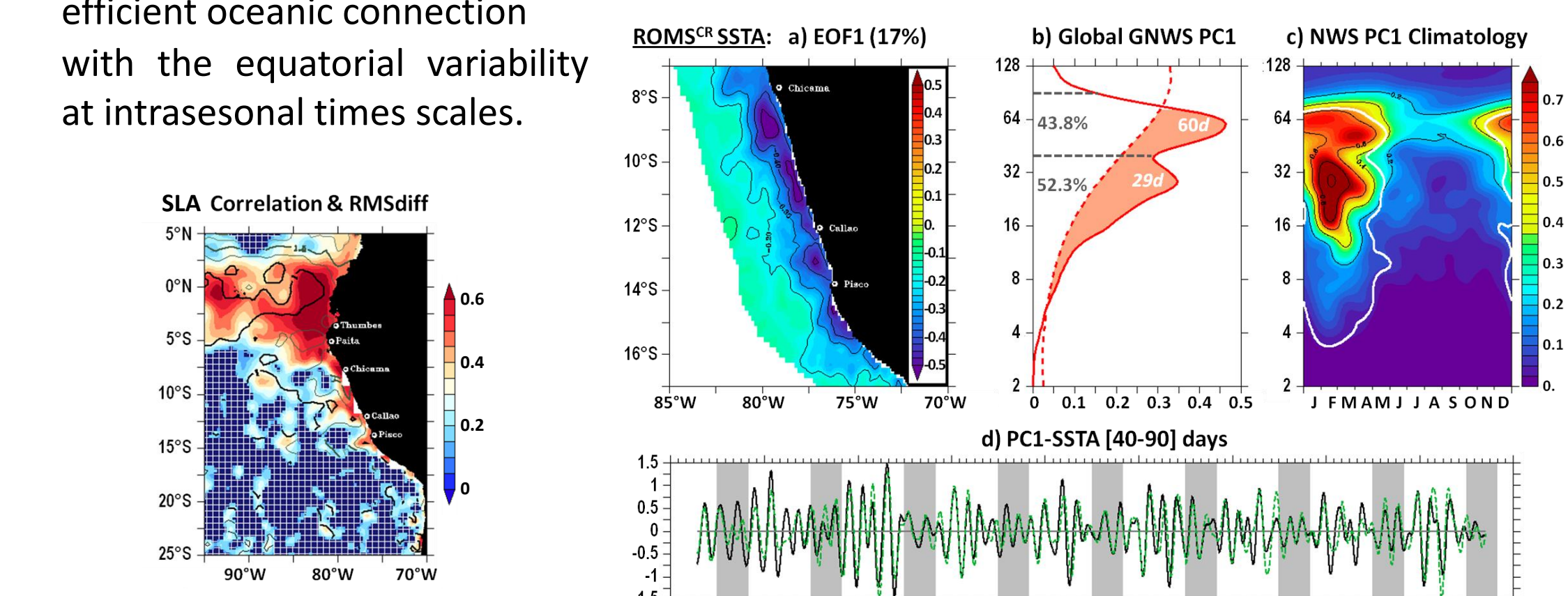


Fig5: Up: same as Fig1 but for ROMS^{CR} SST. Bottom: Intraseasonal band-pass filtered PC1-SSTA (40-90 day⁻¹) for ROMS^{CR} (black) and TMI (green). Grey bands show the July to October season.

Model Experimentation :

EXP-Name	OBC (E)	Wind Stress (W)	Heat Fluxes (Q)
ROMS ^{CR}	Total	Total	Total
ROMS ^E	Total	Climatology	Climatology
ROMS ^{WQ}	Climatology	Total	Total
ROMS ^W	Climatology	Total	Climatology

We carried out 4 experiments in order to assess the oceanic response to wind stress, heat fluxes and long equatorial waves at subseasonal timescales off central Peru and interpret the observations (**Fig1**). ➔ Our experiments differ by their boundary conditions, either climatological or real-time (**Table 1**).

7. References and Acknowledgements

Dewitte, B., S. Illig, L. Renault, K. Goubanova, K. Takahashi, D. Gushchina, K. Mosquera, and S. Purca, 2011: Modes of covariability between sea surface temperature and wind stress intraseasonal anomalies along the coast of Peru from satellite observations (2000–2008), *J. Geophys. Res.*, **116**, C04028, doi:10.1029/2010JC006495.

Goubanova K., S. Illig, E. Machu, V. Gargon and B. Dewitte, 2013: SST subseasonal variability in the Benguela upwelling system as inferred from satellite observations (2000-2008), *J. Geophys. Res.-Oceans*, **2169-9291**, doi: 10.1002/jgrc.20287.

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