

Helmholtz Young Investigator Group VH-NG-1243: "Sub-seasonal PREdictAbility: understanding the role of Diabatic OUTflow" (SPREADOUT)



HELMHOLTZ RESEARCH FOR GRAND CHALLENGES

Development of a logistic model to study warm conveyor belts on sub-seasonal time scales

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WCBs affect lifecycle of blocking and blocking regimes



Pfahl et al. 2015; Steinfeld and Pfahl 2019

Does a misrepresentation of WCBs dilute forecast skill on sub-seasonal time scales?

Systematic verification of WCB in S2S models still missing

Issue 1: Data amount

- ERA-Interim: 40 years * 365 days * 4/day
 - ~58,400 time steps amount to 252 GB trajectory data
- S2S reforecast: 21 years * 66 fc/year * 101 ensemble members * 46 day fc lead time (Vitart et al. 2017)
 - ~6,439,356 time steps would amount to 25 TB trajectory data

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Issue 2: Data availability

• S2S reforecast: Temporal resolution 24-hourly, 11 pressure levels

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Build a statistical model to identify WCBs from Eulerian fields!

- Predictand y: binary fields (0/1 flag) of WCB inflow, ascent and outflow (Madonna et al. 2014; Thanks to ETH Zurich Atmospheric Dynamics group for sharing the data.)
- Predictors $x_1...x_n$ based on ERA-Interim fields of U, V, T, Z, Q on pressure levels available in S2S
- stepwise forward predictor selection using 10x 10-fold cross validation and likelihood-ratio test
- for each grid point on a $5^{\circ}x5^{\circ}$ latitude longitude grid

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Multiple logistic regression model

$$p(WCB|x) = \frac{1}{1 + e^{-g(x)}}$$
 with $0 \le p(x) \le 1$ and $g(x) = \beta_0 + \beta_1 x_1 + \beta_2 x_2 + ... + \beta_n x_n$

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March 2020



WCB inflow Predictors are related to moisture flux

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WCB ascent Predictors are related to moisture flux and thickness advection

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WCB ascent Predictors are related to moisture flux and thickness advection



WCB outflow Predictors are related to relative humidity and irrotational wind speed



WCB inflow

- thickness advection at 700 hPa
- meridional moisture flux at 850 hPa
- moisture flux divergence at 1000 hPa
- moist PV at 500 hPa





WCB inflow

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WCB ascent

- rel. vorticity at 850 hPa
- rel. humidity at 700 hPa
- thickness advection at 300 hPa
- meridional moisture flux at 500 hPa





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WCB ascent

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- rel. humidity at 700 hPa
- thickness advection at 300 hPa
- meridional moisture flux at 500 hPa

WCB outflow

0.005

• irr. wind speed at 300 hPa

Weighted number of occurrences

0.015

0.020

- static stability at 500 hPa
- rel. humidity at 300 hPa

0.010

• rel. vorticity at 300 hPa

Development of one model per grid point and season

Model evaluation - Reliability



Model reliably predicts WCB frequency for probabilites < 0.5 Model overestimates WCB frequency for probabilities > 0.5 (artifact of WCB definition?)

ec.erai = 40 years of training data | ec.erai_20y = 20 years of training data | jra55 = ec.erai applied to jra55 reanalysis

¹⁴ ECMWF Workshop: WCBs – a challenge to forecasting 12 March 2020

Model evaluation - Climatology

- convert predicted probabilities to binary prediction by minimizing climatological bias at each grid point and for each season
- by definition climatology for WCB inflow, ascent and outflow is well reproduced.



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Lagrangian DJF WCB climatology

Model evaluation – Matthews correlation coefficient



- $MCC = \frac{TP \times TN FP \times FN}{\sqrt{(TP + FP)(TP + FN)(TN + FP)(TN + FN)}}$
- MCC=+1 \rightarrow perfect forecast
- MCC=-1 → total disagreement between forecast and observation
- useful for imbalanced data
- high score only if good results for TP, TN, FP, FN





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Case study – January 2011 (Martinez-Alvarado et al. 2016)





Statistical model WCB mask

Case study – January 2011 (Martinez-Alvarado et al. 2016)



Systematic evaluation of WCBs in ECMWF IFS forecasts see next talk by Jan Wandel



Analysed WCB mask

Conclusion

- First attempt of a Eulerian WCB diagnostic via logistic regression model
- Stepwise forward selection identifies most important predictors
 - Inflow: thickness advection, moisture flux, moisture flux convergence
 - Ascent: vorticity, rel. humidity, moisture flux, thickness advection
 - ° Outflow: rel. humidity, divergent wind speed, static stability
- Model skillfully identifies WCB inflow, ascent and outflow footprints

Outlook

- verify WCB footprints in S2S forecast (Jan Wandel)
- finalize WCB diagnostic V2.0 using convolutional neural network
- process studies



WCB ascent

10000km

WCB inflow

WCB outflow

H



Part II: k-fold cross validation







WCB inflow

thickness advection at 700 hPa

• meridional moisture flux at 850 hPa

• moisture flux divergence at 1000 hPa

• moist PV at 500 hPa



WCB ascent

• relative vorticity at 850 hPa

• relative humidity at 700 hPa

• thickness advection at 300 hPa

• meridional moisture flux at 500 hPa



WCB outflow

• relative humidity at 300 hPa

• irrotational wind speed at 300 hPa

• static stability at 500 hPa

• relative vorticity at 300 hPa