HEPEX Earth System modelling @ECMWF – Implications for HEPEX and hydrology

Florian Pappenberger Gianpaolo Balsamo Christel Prudhomme Carlo Buontempo Hannah L. Cloke EFAS@ECMWF

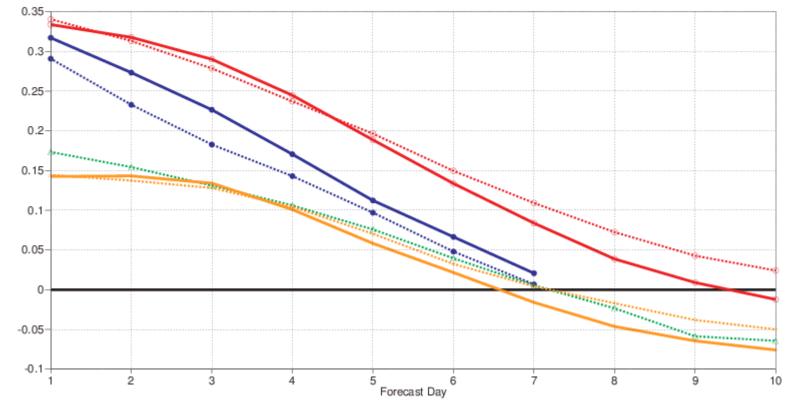
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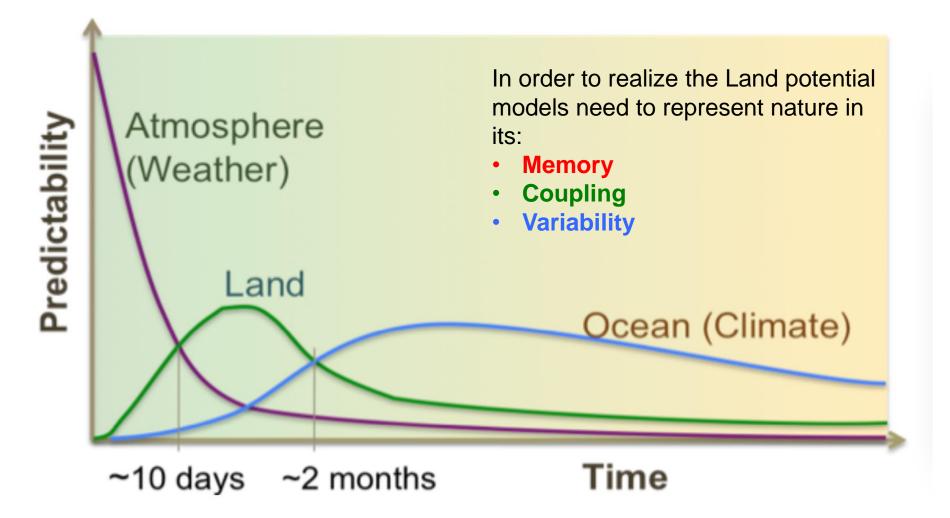
How far can be predict into the future??





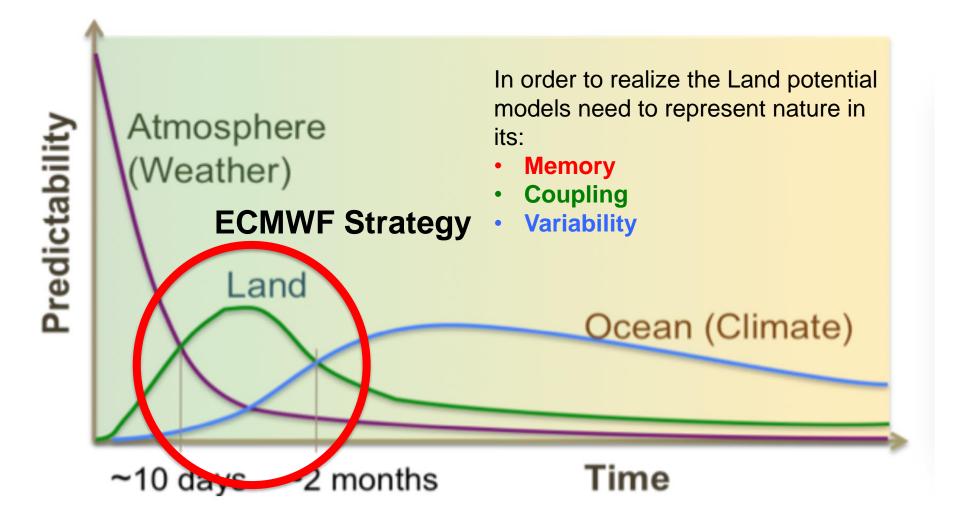
EUROPEAN CENTRE FOR MEDIUM-RANGE WEATHER FORECASTS

Why increased attention on Land and Hydrology by NWP Centres?



Dirmeyer et al. 2015: <u>http://library.wmo.int/pmb_ged/wmo_1156_en.pdf</u>

Why increased attention on Land and Hydrology by NWP Centres?

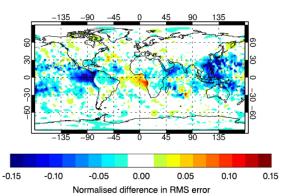


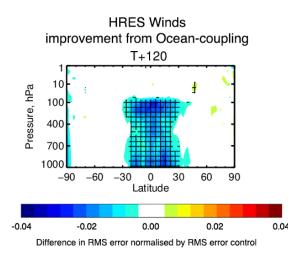
Dirmeyer et al. 2015: <u>http://library.wmo.int/pmb_ged/wmo_1156_en.pdf</u>

On the relative contribution of land and ocean on ECMWF day-5 forecast

Forecast improvements at Day+5 (1 year) Coupled-Ocean vs Uncoupled (only skin-interaction)

HRES Mean Sea-Level Pressure improvement from Ocean-coupling T+120

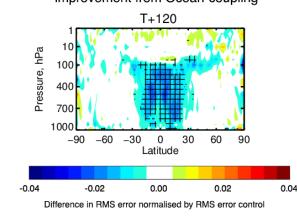




HRES 500 hPa Geopotential Height improvement from Ocean-coupling T+120

-0.1 0.0 0.1 0.2 -0.2 Normalised difference in RMS error

> HRES Relative Humidity improvement from Ocean-coupling



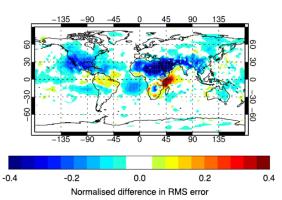
Forecast improvements at Day+5 (1 year) Coupled-Land vs Uncoupled (only skin-interaction)

0.10

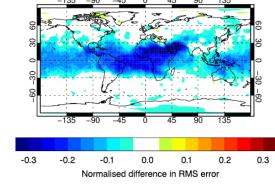
-0.2

TCo399 Mean Sea-Level Pressure sensitivity to Land-coupling T+120

TCo399 500 hPa Geopotential Height sensitivity to Land-coupling T+120



TCO399 Winds sensitivity to Land-coupling T+120 Pressure, hPa 100 400 700 1000 -60 -90 -30 0 30 60 90 Latitude -0.05 0.05 -0.10 0.00 Difference in RMS error normalised by RMS error control



TCO399 Relative Humidity sensitivity to Land-coupling T+120 Pressure, hPa 100 400 700 1000 -90 -60 -30 0 30 60 90 Latitude

0.0 Difference in RMS error normalised by RMS error control

-0.1

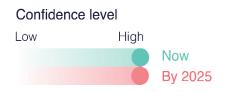
0.1

0.2

Balsamo et al. 2017 WGNE-Blue-book

THINKING AHEAD

How far in advance can we predict extreme weather events, now and in the future?





Earth surface modelling components @ECMWF

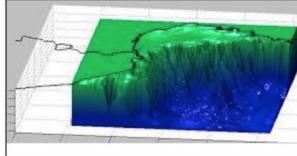
NEMO3.4

NEMO3.4 (Nucleus for European Modelling of the Ocean)

Madec et al. (2008)

Mogensen et al. (2012)

ORCA1_Z42: 1.0° x 1.0° ORCA025 Z75: 0.25° x 0.25°



Hydrology-TESSEL •

R. R.

- Balsamo et al. (2009) van den Hurk and Viterbo (2003)Global Soil Texture (FAO)
- Variable Infiltration capacity & surface runoff revision

EC-WAM

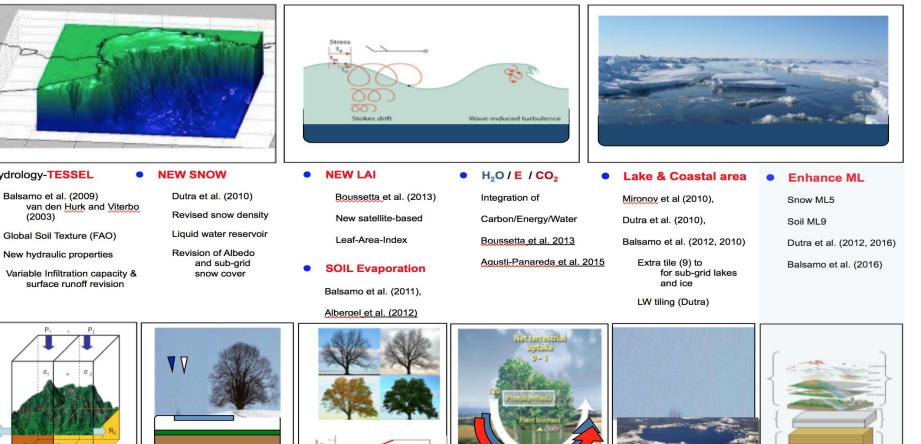
ECMWF Wave Model Janssen, (2004) Janssen et al. (2013)

ENS-WAM : 0.25° x 0.25° HRES-WAM: 0.125° x 0.125°

LIM2 •

The Louvain-la-Neuve Sea Ice Model Fichefet and Morales Magueda (1997) Bouillon et al. (2009) Vancoppenolle et al. (2009)

ORCA025 Z75: 0.25° x 0.25°



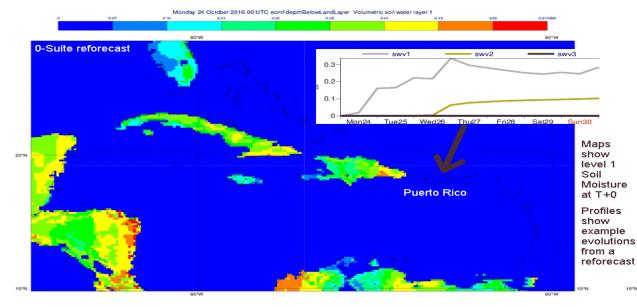
Atm/L and resol.	ECMWF Config. in 2017
80 km	ERAI*
32 km	ERA5* SEAS5
18 km	ENS
9 km	HRES ⁺

Ocean 3D-Model Surface Waves and currents, Sea-ice. *(ocean-uncoupled) +(coupled in 2018)

Land surface 1D-model soil, snow, vegetation, lakes and coastal water (thermodynamics). Same resol. as Atm.

ERA5-Land: a 9-km land reanalysis (high-resolution downscaling, water conserving)

ERA5/Land benefits from the R&D done in EartH2Observe H2020 project and Copernicus, is based on the new ERA5 and it run at 9-times higher resolution than ERA-Interim and 3.5-times higher resolution than ERA5. ERA5/Land will match the highest resolution currently operational at ECMWF (HRES 9km)



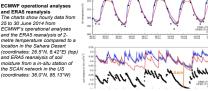
Suite reforecast Puerto Rico

Munoz-Sabater et al on ERA5-Land description and performance



Support hydrological studies addressing global water resources Provide consistent land initial conditions to weather and climate models Foster research into intra-seasonal forecasting

Provide dedicated datasets to support and encourage downstream land applications



Dutra et al on high resolution downscaling and performance

A temporally and spatially varying environmental lapse-rate for temperature downscaling

E. Dutra⁽¹⁾, J. Muñoz-Sabater⁽²⁾, S. Boussetta⁽²⁾, T. Komori⁽³⁾, S Hirahara⁽³⁾, and G. Balsamo⁽²⁾ ⁽¹⁾ Instituto Dom Luiz, Faculdade de Ciências, Universidade de Lisboa, Portugal; ⁽²⁾ European Centre for Medium-Range Weather Forecasts, United Kingdom; ⁽³⁾ Global Environment and Marine Department, Japan Meteorological Agency. Corresponding author: endutra@fc.ul.pt

Motivation

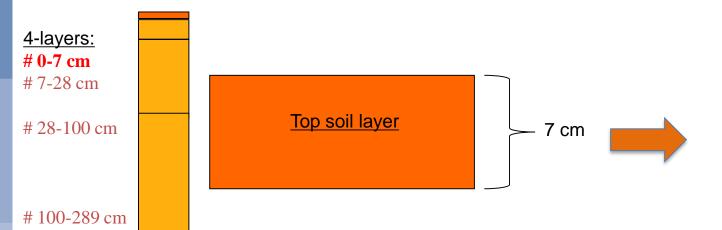
Temperature near the surface varies with altitude accordingly to th environmental lapse-rate (ELR). The ELR depends on the overlying a masses, large-scale situation and local effects. In this study we propose th derivation of the ELR from the reanalysis lower troposphere vertic profiles of temperature. This creates a temporally and spatially varying ELR, that can be used to downscale near-surface air temperature from th reanalysis resolution to higher resolutions.

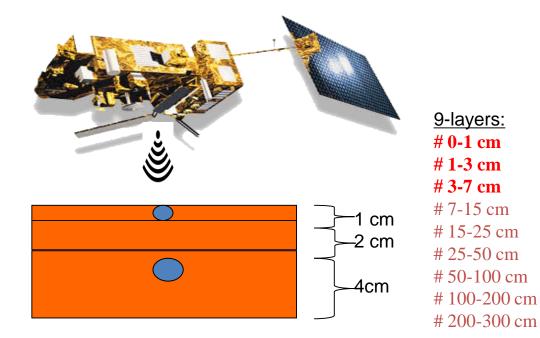
	Table 1.	
he air he cal ng he	Acronym	Details
	E5	ERA5 reanalysis (35 km)
	CLR	Direct downscaling to station elevation using a constant ELR of -6.5 K $\rm km^{-1}$
	MLR	Direct downscaling to station elevation using a climatological EL
	DLR	Direct downscaling to station elevation using a daily ELR
	bil5	Surface only at 9 km driven by E5 with bilinear interpolation
	clr5	As bil5 but with a constant ELR of -6.5 K km ⁻¹ temperature

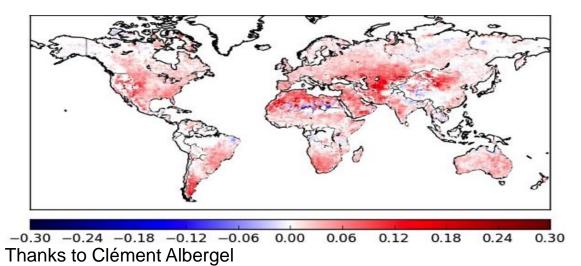
Thanks to Joaquin Munoz-Sabater and ERA-Team

Future enhanced soil model vertical resolution to increase use of satellite data

The model bias in Tskin amplitude shown by <u>*Trigo et al. (2015)*</u> motivated the development of an enhanced soil vertical discretisation to improve the match with satellite products.







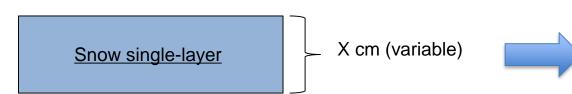
Comparison with ESA-CCI soil moisture remote sensing (multi-sensor) product.(1988-2014). A finer soil model improves the correlation with measured satellite soil moisture

Globally Improved match to satellite soil moisture (shown is Anomaly correlation ΔACC calculate on 1-month running mean)

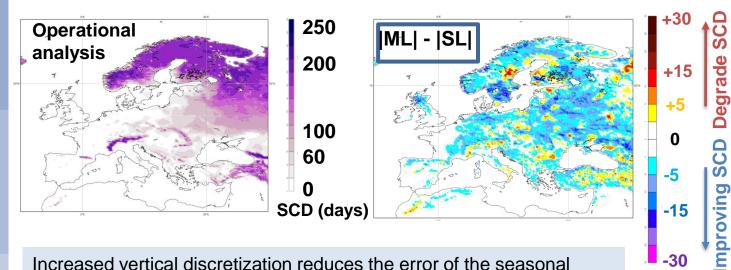
Future enhanced snow model vertical resolution: impact in cold regions climate

Increased vertical discretization of the snowpack (up to 5 layers) permits a better physical processes representation

-30

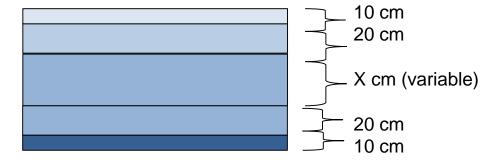


Snow cover duration (SCD) over Europe 2016/2017

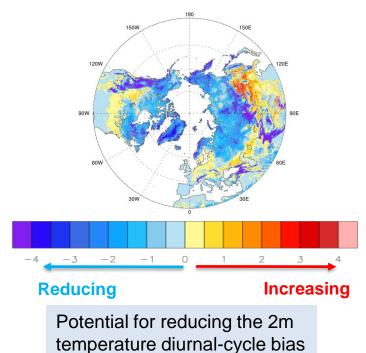


Increased vertical discretization reduces the error of the seasonal snow cover duration by 5 to 15 days (evaluated in ERA5-Land mode)

Thanks to Gabriele Arduini, Souhail Boussetta, Emanuel Dutra



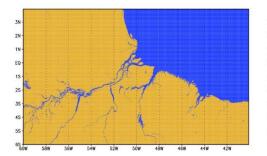
Difference in T_{skin} minimum winter (DJF)



Mapping the surface at 1km: water bodies and changes over time

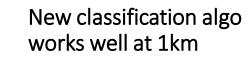
Classifying automatically inland water bodies is a complex task. A 1-km water bodies cover and bathymetry have been produced

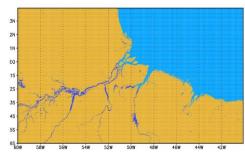
ESA GlobCOVER has no water class



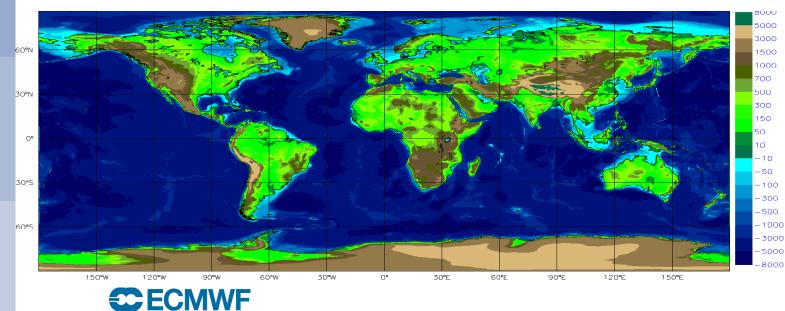


Flooding allows classify,





A new 1-km global bathymetry and orography map (SRTM+/GEBCO/GLDB)



Thanks to Margarita Choulga, Souhail Boussetta, Irina Sandu, Nils Wedi

ESA GlobCOVER is combined with JRC/GLCS to detect Lake cover changes



ECMWF Newsletter No. 150 - Winter 2016/17

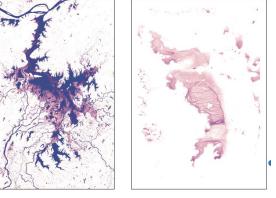
Lakes in weather prediction: a moving target

GIANPAOLO BALSAMO (ECMWF), ALAN BELWARD (Joint Research Centre)

Lakes are important for numerical weather prediction (NWP) because they influence the local weather and climate. That is why in May 2015 ECMWF implemented a simple but effective interactive lake model to represent the water temperature and lake ice of all the world's major inland water bodies in the Integrated Forecasting System (IFS). The model is based on the version of the FLake parametrization developed at the German National Meteorological Service (DWD), which uses a static dataset to represent the extent and bathymetry of the world's lakes.

However, new data obtained from satellites show that the world's surface water bodies are far from static. By analysing more than 3 million satellite images collected between 1984 and 2015 by the USGS/NASA Landsat satellite programme, new global maps of surface water occurrence and change with a 30-metre resolution have been produced. These provide a globally consistent view of one of our planet's most vital resources, and they make it possible to measure where the world's surface water bodies really can be found at any given time.

As explained in a recent Nature article (doi:10.1038/nature20584), the maps show that over the past three decades almost $90,000 \text{ km}^2$ of the lakes and rivers thought of as permanent have vanished from the Earth's surface. That is equivalent to Europe losing half of its lakes. The losses are linked to drought



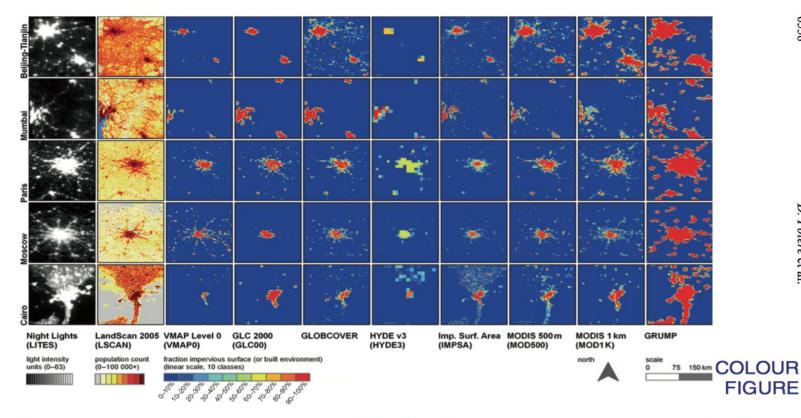
Dynamic lakes. The size of Poyang Lake (left), one of China's largest lakes, fluctuates dramatically between wet and dry seasons each year while overall decreasing. Lake Gairdner in Australia (right), which is over 150 km long, is an ephemeral lake resulting from episodic inundations. Both maps show the occurrence of water over the past 32 years: the lighter the tone the lower the occurrence. (Images: Joint Research Centre/Google 2016)

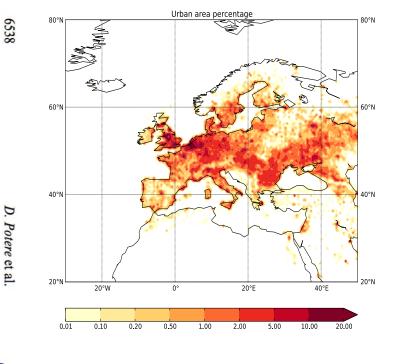


Lake Victoria. Lakes in tropical areas are linked with high-impact weather by contributing to the formation of convective cells. (Photo: MHGALLERY/iStock/Thinkstock)

Mapping the surface at 1km: urban cover, its expansion and uncertainties

Classifying automatically urban areas (fraction and height) is an extremely complex task. Urban areas expand each year





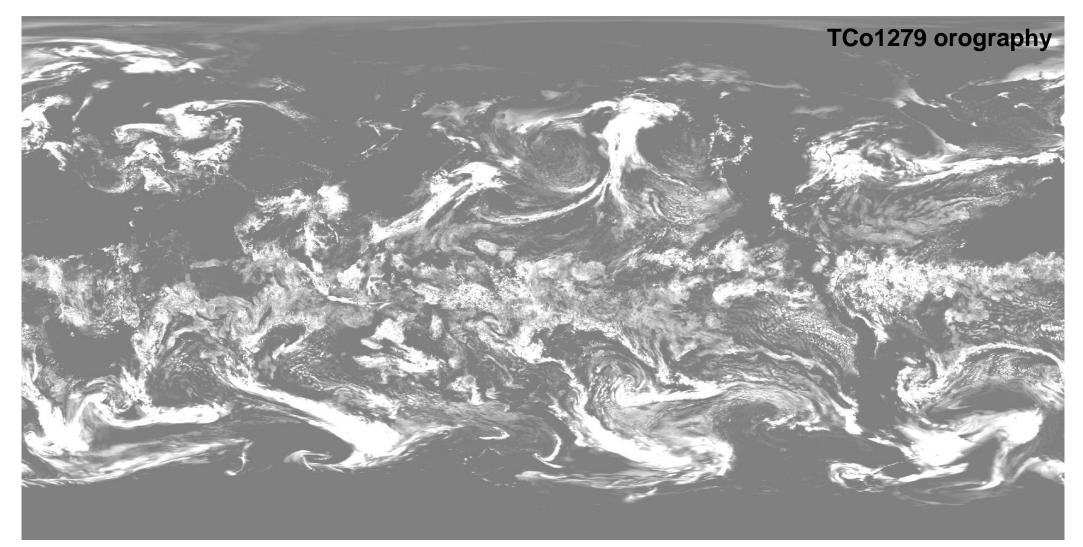
- Urban area (a, in %, from ECOCLIMAP Masson et al., 2003) see:
- Balsamo et al. 2014 ECMWF TM729
- CHE H2020 Project <u>http://www.che-project.eu</u>

Figure 1. The eight global urban maps and two urban-related maps for Beijing-Tianjin, China (top row), Mumbai, India (second row), Paris, France (third row), Moscow, Russia (fourth row), and Cairo, Egypt (bottom row). LITES, LSCAN and IMPSA are at native 30 arc-second resolution, HYDE3 is at native 5 arc-minutes, and the remaining maps have been aggregated from 30 arc-seconds to 1.5 arc-minutes for display. This aggregation effectively converts their legends from binary (urban/rural) to continuous (percentage urban).

 Urban area dataset comparison on selected cities (Potere et al., 2009 IJRS) reveal large uncertainties and discrepancies

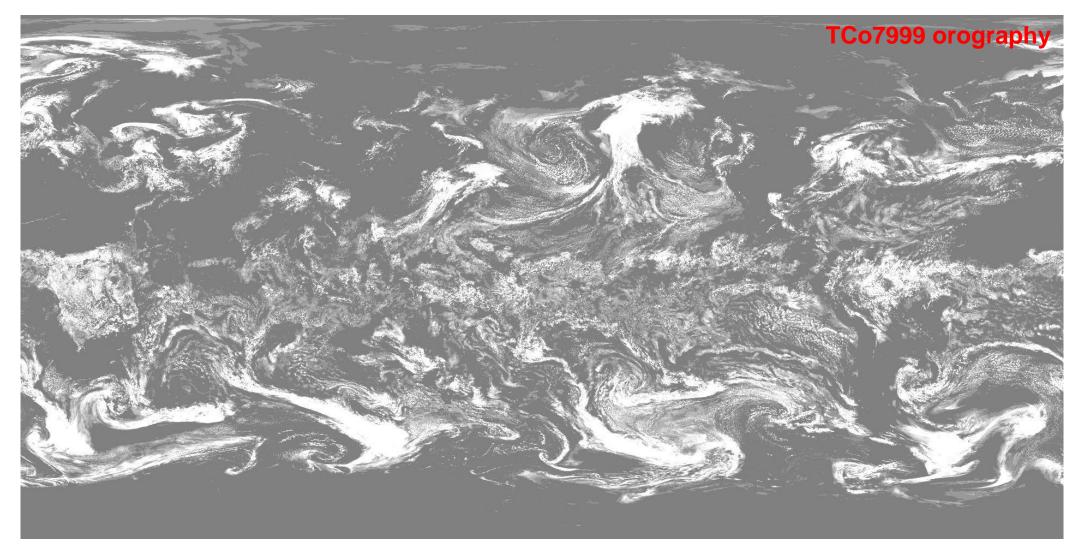


Current km-scale: TCo1279 (~9km) highest global operational NWP today



(12h forecast, *hydrostatic*, with deep convection parametrization, 450s time-step, 240 Broadwell nodes, ~0.75s per timestep) **EQUIVALENT to 6.6 Megapixel camera**

Towards km-scale: TCo7999 test-case (~1.3km) highest NWP test @ECMWF



(12 h forecast, *hydrostatic*, no deep convection parametrization, 120s time-step, 960 Broadwell nodes, ~6s per timestep in SP)

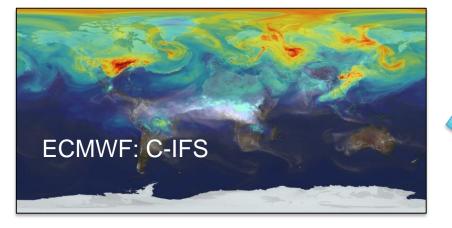
Thanks to Nils Wedi and NM-Team

Equivalent to 256 Megapixel camera

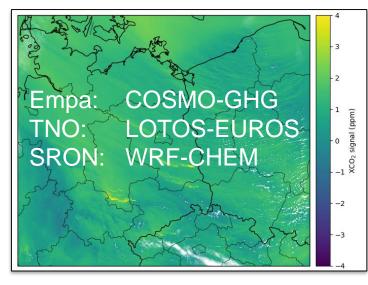


Embracing multi-scale to evaluate uncertainties, from local to global

Global, ~ 9km resolution, ECMWF



Regional, ~ 1 km, Empa, TNO, SRON

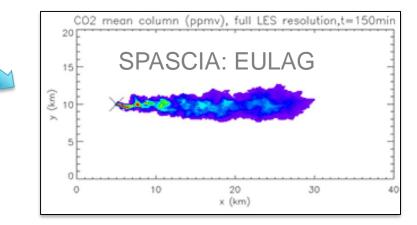


CECMWF

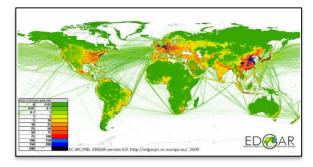
Europe, ~ 5 km, Empa, TNO, MPG



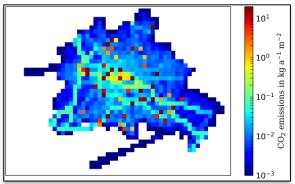
Point source, ~ 100 m, SPASCIA



Global, Regional City emission







Land surface uncertainty – LAND SURFACE – ATMOSPHERE FEEDBACKS



Improved seasonal prediction of the hot summer of 2003 over Europe through better representation of uncertainty in the land surface

David A. MacLeod, ^a* Hannah L. Cloke, ^{b,c} Florian Pappenberger^d and Antje Weisheimer^{d,e} ^aAtmospheric, Oceanic and Planetary Physcis, Department of Physics, University of Oxford, UK ^bDepartment of Geography and Environmental Science, University of Reading, UK ^cDepartment of Meteorology, University of Reading, UK ^dEuropean Centre for Medium-Range Weather Forecasts, Reading, UK ^eDepartment of Physics, National Centre for Atmospheric Science (NCAS), University of Oxford, UK

*Correspondence to: D. A. MacLeod, Department of Physics, Clarendon Lab, Parks Road, Oxford, OX1 3PU, UK. E-mail: macleod@atm.ox.ac.uk

Methods to represent uncertainties in weather and climate models explicitly have been developed and refined over the past decade and have reduced biases and improved forecast skill when implemented in the atmospheric component of models. These methods have not yet been applied to the land-surface component of models. Since the land surface is strongly coupled to the atmospheric state at certain times and in certain places (such as the European summer of 2003), improvements in the representation of land-surface uncertainty may potentially lead to improvements in atmospheric forecasts for such events.

Here we analyze seasonal retrospective forecasts for 1981–2012 performed with the European Centre for Medium-Range Weather Forecasts (ECMWF) coupled ensemble forecast model. We consider two methods of incorporating uncertainty into the landsurface model (H-TESSEL): stochastic perturbation of tendencies and static perturbation of tendencies.

Hydrol. Earth Syst. Sci., 20, 2737–2743, 2016 www.hydrol-earth-syst-sci.net/20/2737/2016/ doi:10.5194/hess-20-2737-2016 @ Author(s) 2016. CC Attribution 3.0 License.



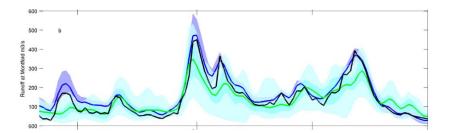
Evaluating uncertainty in estimates of soil moisture memory with a reverse ensemble approach

Dave MacLeod¹, Hannah Cloke^{2,3}, Florian Pappenberger^{4,5}, and Antje Weisheimer^{4,6}

¹Atmospheric, Oceanic and Planetary Physics, Department of Physics, University of Oxford, Oxford, UK
²Department of Geography and Environmental Science, University of Reading, Reading, UK
³Department of Meteorology, University of Reading, Reading, UK
⁴European Centre for Medium-Range Weather Forecasts, Reading, UK
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Received: 18 January 2016 – Published in Hydrol. Earth Syst. Sci. Discuss.: 17 February 2016 Accepted: 6 June 2016 – Published: 12 July 2016

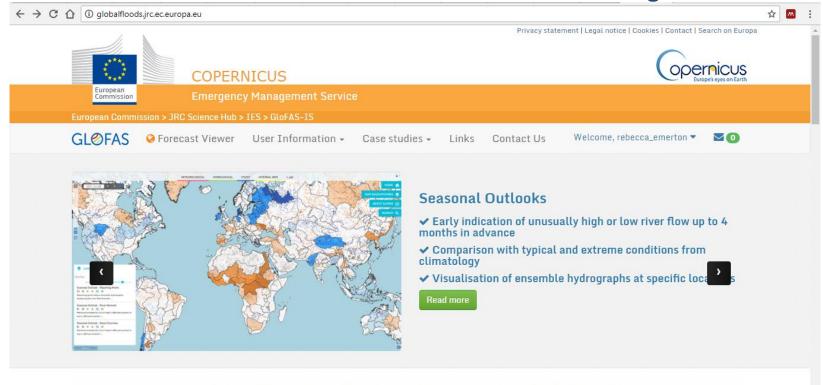




Abstract. Soil moisture memory is a key component of seasonal predictability. However, uncertainty in current mem-

GloFAS-Seasonal: Forecast Products

www.globalfloods.eu



It couples state-of-the art weather forecasts with an hydrological model. Find out more!

Emerton et al (in prep) & here at HEPEX!





WHAT WILL THE INFORMATION BE USED FOR?

The wealth of climate information will be the basis for generating a wide variety of climate indicators aimed at supporting adaptation and mitigation policies in Europe in a number of sectors. These include, but are not limited to, the following:



- to provide practical examples of how C3S in general and CDS in particular could deliver information of relevance to specific sectors.
- 2) To provide examples of good practice. This means that the SISs should be built to the highest possible standards so that services developers could be inspired by them and look at them as quality benchmarks.
- 3) To provide information on users needs, and whenever possible address those. In particular SIS contract should develop and make available sector-relevant indicators and tools that were either unavailable or inaccessible before.



EartH2Observe H2O20 multi-model GHP/LSMs ensemble and its added value

Multi-model mean/median outperforms single models for most of the variables

HTESSEL-CaMa WRR2-MEDIAN WRR2-MEAN EAFHYDRC LISFLOOD SURFEX-TR PCR-GLOB W3-v2 Wat Latent Heat Spatial distribution score of latent heat Evapotranspiration 2. 0^{.0} 0.1 0.2 0.3 Surface Soil Moisture 1.8 Correlation 1.6 Soil Moisture Anomaly 1.4 *.*~ 1.2 Snow Water Equivalent 1.0 6.0 0.8 Snow Cover 6.95 0.6 **Terrestrial Water Storage** 0.4 0.99 Anomaly 0.2 0. 0 ⊾ 0. 0 Runoff 0.8 1.0 1 2 1.4 1.6 1.8 2.0 0.2 0.4 Ο. 6 Normalized standard deviation 0.25 0.5 0.75 -2 -1 +0+2 1 +10 LISFLOOD W3-v2 Benchmark Variable Score Variable Z-score HTESSEL-CaMa ORCHIDEE ergar JULES PCR-GLOBWB WRR2-MEAN Thanks to Alberto Martinez, Eleanor Blyth, and E2O-Team WRR2-MEDIAN LEAFHYDRO SURFEX-TRIP

ESSEL-CaMa

FHYDRO

HIDFF

8

Conclusions

• Land surface schemes are getting very advanced – high resolution e.g. 1km will happen (test runs are here, operational next)

- We are working towards a coupled and closed water cycle
- Uncertainty representation is a challenge in an operational environment (amount of data produced only allows for a limited number of ensembles) – but we can demonstrate that we can do it
- Earth System modelling approach is a viable way of improving weather forecasts (and delivers environmental forecasts at the same time)



Wildfire App; Webcrawler for Hydrological data, data extraction tool, innovative visualisation, Machine Learning (optimize data requests from MARS archive) and many more



EUROPEAN CENTRE FOR MEDIUM-RANGE WEATHER FORECASTS

http://esowc.ecmwf.int/