6-P01 On the causes of poleward shift of the Indian summer monsoon low level jetstream

Krishna AchutaRao, S. Sandeep, Dileepkumar R, and Arulalan T Centre for Atmospheric Sciences, Indian Institute of Technology Delhi



Contrasting methods of detecting and attributing the impact of external forcings Session 6

Julie Arblaster (julie.arblaster@monash.edu) and Catalyst members

Poster 2

ENERGY

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catalyst

Attribution of Ocean Temperature Change to Anthropogenic and Natural Forcings using the Temporal, Vertical and Geographical Structure







Roberto Bilbao (roberto.bilbao@bsc.es), Jonathan Gregory, Nathaelle Bouttes, Matthew Palmer and Peter Stott

Comparison of Ocean temperature Observations and CMIP5 models (1960-2005):

Observations and CMIP5 models show that the upper 2000m has warmed with a signal that has a well defined geographical pattern and vertical structure.

Greenhouse gas forcing has contributed most to increasing the temperature of the ocean, a warming which has been o set by other anthropogenic forcing (mainly aerosols), and volcanic eruptions which cause episodic cooling.

Detection and attribution analysis:

We carry out multi-model detection and attribution analysis, using optimal fingerprinting, based on the time and depth structure of the temperature together:

- Two-signal: anthropogenic and natural forcings.
- Three-signal: greenhouse gas, anthropogenic aerosols and natural forcings.







Figure 2. ANT (blue) and NAT (green) signals scaling factors and uncertainty ranges for global mean ocean temperature change between 1960-2005 for multiple depth level fingerprints.

Implication of Mid Holocene and Last Interglacial changes in insolation seasonality on high and mid latitude climate

P. Braconnot, B. Otto-Bliesner and PMIP participants

Two interglacial periods : changes with modern conditions driven by insolation



- Understand radiative forcing/atmospheric and ocean circulation
- Seasonality and atmosphere, ocean, landsurface and cryosphere feedbacks
- Model benchmarking against paleoclimate reconstructions
- * Test model evolution through time (PMIP1 to PMIP4)

Similarities and differences in the response of the NH high latitude climate



Discuss also hydrology, sensitivity test to vegetation, dust.....

Specific thanks to Paolo Scussolini, Marie Sicard, Jérôme Servonnat and Jean-Yves Peterschmitt



THE RESPONSE OF CLIMATE VARIABILITY IN PMIP4/CMIP6

Chris Brierley¹, Julia Hargreaves², Kira Rehfeld³ on behalf of the PMIP4 P2FVAR working group

ersity College London: ²Blues Skies Research: ³U

VARIABILITY

INTRODUCTION

The Deleanclimate Model Intern Project (PMIP4) consists of Mid-Holocene /6 000 years and warmish's Last glacial maximum (21,000 yrs ago; cold) Last interglacial (127,000 years ago; warm) Pliocene (-3,000,000 years ago; -400 ppm)

 Last Mileonium (transient, past 1,000 years) nal sensitivity and PMIP-only expts

The 'Past2Future: insights from a constantly varving past" (n2har) working group seeks to help climate scientists make the best use of the fact that PMIP4/CMIP6 includes many models and time periods. We host a database of Climate Variability Diagnostics Package (CVDP, Phillips et al., 2014) output. Data and scripts used on this poster can be found at cl.ac.uk/~ucfaccb/PMIPVarData

Made	Nid- Balacence	Last Glacial Ogni		Allenge	Intergladial Rig1276	of tempe
ENSO	0,78	1.03	1.9	1.72	0.46	0
(Niñe 3.4)	0.87		2.37	2.15		8
Tropical Indian	1.64	1.34	1.2	1.52	1.17	10
Occus 55T	0.96		1.60	1.70		-
Jobal Mean Surface	1.55	1.48	0.75	0.9	0.65	E .
Air Temperature	1.05		6.69	0.79		0
Southern Ocean	1	0.51	1.16	0.95	0.67	E
SST	9.97		0.40	0.66		0
Tropical South	0.93	1.28	1.03	0.84	1.07	5
Atlantic SST	1.11		1.09	0.97		8
Tropical North	0.91	1.33	1.01	1.26	1.05	26
Atlantic SST	1.08		12.96	1.12		č
Indian Ocean Dipole	1,04	1.51	1.2	0.95	1.89	w
(DMD	2.28		1.59	1.16		
Atlantic Multidecadal	1.11	1.67	1	1.16	1.07	
Oscillation	1.83		4.85	0.87		
Atlantic Meridianal	0.87	1.53	0.79	0.65	0.97	
Mode	1.01		0.59	0.58		Fig. 2
Atlantic Nilo	1.05	1.8	0.58	0.49	0.93	
(#103)	1.06		0.58	0.50		
Pacific Decadal	1.62	1.21	1.25	1.36	1.22	0
Oscillation	1.10		2.04	0.94		<u>0</u>
Intendecedal Pacific	1.31	1.72	6.4	4.44	1.15	5
Oscillation	1.26		5.77	2.77		ĉ
Northern Annular	0.97	0.9	1.13	1.11	0.98	6
Mode	0.05		1.19	1.12		1
North Atlantic	1.03	0.82		1.26	0.97	5
Oscillation	1.07		1.16	2.26		- P
Southern Annular	1.06	1.06	1.11	0.91	1.13	2
Made	1.04		0.91	0.91		E.
mplitude of varios my CMIP6 models. his is the percent. Marwise, it shows to occur-sported areas soccured with the oc	for inc ge ch he perc serage serage	ante m les def ange i entage of t	odes, inol n n ind chuoy he E cipal o	Italies podes jex en ge in d QF p comps	in the slow (bold), nance, ne root sattern nent	ratio of pre-
Figs 1 & 2: Ensemi Percentage charge imperature (Fig. 1, is are first computed in across all membs	ble me in local (p) and reach a crs of 4	an chu Innui pecip specip CMIP5	inges I meat itation itation itation itation itation itation itation	in sur voria v (Fig. hen av	iance. nce of 2, k(R) eraged ance is	semble mear
only the final 30 yr Silppid adore % of	ars of a	the tran	vient	capen	ments,	μ.

PAST 2 FUTURE Analysis focuses around either climate modes The past provides a testbed to evaluate future (Table) or, more recently, local variance. projections, and methods now exist to make Rehfeld et al. (2018) show large increases in uantitative inferences from comparisons with palaeobservations. Previous work has local variance during the cold glacial. Cox et al. (2018) claim it constrains climate sensitivity. We show ensemble-mean variance changes in constrained climate sensitivity (Hargroaves & Annan, 2016) and hydroclimate (Saint Lu ef temperature (Fig. 1) and precipitation (Fig. 2). al. 2015) Below we show examples when ch methods may, and may not, be applie

AICI



PMIP4 P2FVAR wg

- PMIP4 provides a useful testbed for research into future climates, by combining models with palaeoclimate data.
- Working group exists to help scientists make the best use of the fact that PMIP4/CMIP6 includes many models and time periods.
- We host a database of pre-processed simulation output & some scripts to analyse it.
- Models with high climate sensitivity running • past warm climates would be rather useful

Session 6, poster 6 **Increased variability of eastern Pacific El Niño under greenhouse warming** Non-selected, 1900-2099 DJF Cai *et al.* (2018) a Observed first principal pattern 0.025 Nm⁻¹ Modelled relationship 0.1 20⁰N С -0.1 $\alpha_{\rm obs}/2$ 20°S eudly -0.2 -0.3 Selected models, 1900-2099 DJF ponent (s.d.) 120°W 180°W 60°W α_{obs} -0.4 b Observed second principal pattern 0.025 Nm⁻ Corre.Coeff.= -0.92 Corre.Coeff.= 0.84 Slope= 0.43 -0.5 Slope= -0.24 8 p<0.001 p<0.001 20°N Second principal -0.6 -0.6 -0.3 0.3 0.6 0.9 1.2 0.3 1.5 -0.9 -0.6 -0.3 Skewness of E-index Skewness of C-index 20°S -4 -2 0 2 First principal component (s.d.) 120^oE 120^oW 180°W 60°W Modelled EP patterns Projected changes in variability of monthly E-index c ORA-s4 SST, 1979-2010 DJF Control 10N Climate change



0

CESM1-CAM5

IPSL-CM5A-LR

150E

10S

10N 0

10S



Climate Scenarios for the Fifth United States National Climate Assessment David R. Easterling¹, Kenneth E. Kunkel², and Xungang Yin³

- Hurricane Harvey (2017) dropped more than 1200mm of precipitation on the Houston, Texas area over a 5 day period.
- How well do climate models produce these kinds of multi-day heavy precipitation events compared to obs?
- Model Simulations.

CMIP6: NOAA GFDL CM4 model, PiCtrl CMIP6: IPSL model, PiCtrl and Doubled CO2 CMIP5: NOAA GFDL CM3, Historical and PiCtrl

Results:

- The GFDL CM4 models results are superior to GFDL CM3 model results in event magnitude, although the seasonal distribution is biased and events are too large at the 100,000 km2 scale
- The IPSL model events are a little higher than observed when comparing similar box sizes
- At their native resolutions, none of the model simulations produce an event of the size of Harvey





Transient Climate Response to Cumulative Emissions in CMIP6 models

Preliminary results from the C4MIP experiments

Pierre Friedlingstein, Chris Jones, Vivek Arora, Tatiana Ilyina and the C^4 MIP community

I) Background TCRE is a metric that measure the global average surface IPCC AR5 SYF

temperature change for a given cumulative CO_2 emissions. IPCC AR5 assessed that TCRE range is 0.8 to 2.5°C/10³GtC. TCRE gained a large interest in the policy arena as it allows to quantify the remaining carbon budget for a given climate target, with a large TCRE implying a low remaining carbon budget.

CMIP6 provides an opportunity to reassess TCRE with state of the art Earth System Models (ESMs). The deck 1% simulation allows to quantify TCRE providing ESMs simulate land and ocean carbon sinks, anthropogenic CO₂ emissions being diagnosed as: $E(t) = \frac{dC_A}{dt} + F_{land} + F_{ocean}$ (1)

Formally, TCRE can be expressed as the product of a measure of the climate sensitivity by the atmospheric CO₂ airborne fraction:

$$TCRE = \frac{dT}{dE_{cum}} = \frac{dT}{dC_A} \times \frac{dC_A}{dE_{cum}}$$
(2)

In addition the C⁴MIP 1% BGC and 1% RAD simulations allow to quantify the strength of the carbon cycle feedbacks (β and γ) and their contribution to the TCRE uncertainty.

2) TCRE in CMIP6



A preliminary analysis based on four ESMs available to date, CNRM-ESM2-1, MPIESM, IPSL-CM6-ESM, and UKESM is presented here. The spread in TCRE is already quite large, with diagnosed TCRE of 1.4° (CNRM-ESM2-1), 1.6 MPIESM), 1.9 (IPSL-CM6-ESM), and 2.3 °C/103GtC (UKESM)., although still within the IPCC AR5 assessed range (0.8-2.5 °C/103GtC).

From the models available so far, the spread in TCRE largely comes from the spread in climate sensitivity, (CS), CNRM-ESM2-1, MPIESM, IPSL-CM6-ESM and UKESM having respectively a Transient Climate Response (TCR) of 2.8, 2.8, 3.5°C and 4.2 °C.

However, the uncertainty in the airborne fraction (AF) is not negligible, CNRM-ESM2-I lower TCRE than MPIESM being primarily due to its slightly lower airborne fraction.

3) Carbon Cycle Feedbacks



Further analysis of the carbon cycle role in controlling TCRE via the airborne fraction could be done using the simple linear climate-carbon feedback framework. TCRE can be expressed as:

$$TCRE = \alpha / (1 + \beta + \alpha \gamma)$$

with α being a measure of the climate sensitivity, β being the carbon cycle sensitivity to atmospheric CO₂ increase, and γ being the carbon cycle sensitivity to climate change.

(3)

As in CMIP5, models show a large spread in land carbon cycle response to both atmospheric CO_2 and climate, while the ocean carbon cycle response is more robust across the models available here.

Spread in land response could be due to presence/absence of nitrogen cycle.

4) (very preliminary) Conclusions



- TCRE can be diagnosed from 1%CO2 runs performed by CMIP6 ESMs.
- From the models available now, all have a TCRE above the CMIP5 multi-model mean, with 2 models being above the $I-\sigma$ range of the CMIP5 ESMs.
- Nevertheless, all models are still within the AR5 assessed range (0.8-2.5 °C/103GtC).
- Preliminary analysis indicate that the large TCRE simulated by IPSL-CM6-ESM and UKESM is primarily due to the large climate sensitivity of these models.
- Spread in land carbon cycle response to CO_2 and climate is quite large, potentially due to nitrogen cycle being only included in some ESMs.



Assessing the robustness of marine heatwave projections

Thomas Frölicher (froelicher@climate.unibe.ch), Mathias Aschwanden, Stephen Griffies



MHWs have occurred in all ocean basins over the last decades.



The number of MHW days have doubled since 1982. 87 % of today's MHWs have an anthropogenic component.





Largest changes are projected for tropics and Arctic Ocean. Changes are mainly driven by global-scale shift in mean SST.

UNIVERSITÄT

6-PIO: Sensitivity of precipitation and its future changes to model resolution Ying Na and Qiang Fu

Dept. of Atmos. Sciences, University of Washington, Seattle, WA, USA

Motivation: Examining how the precipitation probability density functions, extreme precipitations, and clear-sky fractions, and their future changes depend on the model resolutions

Data: IPSL-CM6A-LR (1950-2014) and the high resolution cloud resolving model NICAM (1979-2008) and 4 scenarios of IPSL-CM6A-LR (2015-2100) and NICAM based on A1B scenario (2075-2104)

Main Conclusion: For a given model, the frequency of extreme precipitation and clear sky fraction tend to decrease but their future changes tend to increase when the model data are re-gridded to coarser resolution



Fig. (a) Precipitation probability density of IPSL and NICAM with native resolution and coarser resolution for historical (1979-2008) and future (2071-2100 for IPSL, 2075-2104 for NICAM). (b) Percentage change of precipitation probability scaled by temperature change. The values under the labels are clear sky fractions and their percentage changes in the future scaled by temperature change.

Evaluation of the PMIP4/CMIP6 palaeosimulations: P11

MAR









New data syntheses, Improved theoretical basis, Forward models, **Better evaluation tools**



mean upward motion is weakened over

the monsoon regions, with a less

weakening over Asia.

Monsoon precipitation responses to global warming and their regional differences simulated by CMIP models



• The second factor is the most influential in the South Asian monsoon, resulting in the largest increase in precipitation, suggesting an important role of the land warming on the Asian monsoon response.

6_P13_ilyina_tatiana

How far is the carbon sink predictable in a multi-model framework?

Ocean C sink: predictable up to 2-3 years globally and up to 6 years regionally **Land C sink**: predictable up to 2 years primarily in the tropics and extra-tropics



29 March – 9:30-15:00 Carbon Cycle Predictability Meeting Venue: Aula de Teleensenyament – B3 Building, 1st floor



Tatiana Ilyina, H. Li, A. Spring, R. Bernardello, L. Bopp, J. Dunne, P. Friedlingstein, N. Lovenduski, M. Chikamoto, J. Park, R. Séférian, S. Yeager



Preliminary results from the Global Carbon Cycle emissions of driven simulations in the NASA-GISS climate model

CMIP6 Workshop Session #6 Poster P14

Gen Ito, Anastasia Romanou, Nancy Kiang, Igor Aleinov, Gregory Faluvegi, Maxwell Kelley, and Reto Ruedy

NASA Goddard Institute for Space Studies, New York, NY, USA (contact gen.ito@nasa.gov)

NASA GISS ModelE 2.1 coupled land-ocean-atmosphere global simulation run in:

- 1. concentration-driven historical simulation for 1850-2015 forced by prescribed CO₂
- emissions-driven simulation forced by anthropogenic CO₂ emissions interacting with the atmosphere and coupled to the model's radiation



Fully coupled emissionsdriven simulation consistent with the concentrationdriven case:

- atmospheric CO₂
- land/ocean fluxes

Perform all tier 1 experiments described in C4MIP protocol

Transient simulations over the Common Era as part of PMIP4/CMIP6

Johann Jungclaus¹, Alexandra Jahn², Matthew Toohey³, Sebastian Wagner⁴, and Stephan Lorenz¹



Northern Hemisphere sea ice NH SIA SEPT $[10^{-12} \text{ m}^{2}]$ 300 600 900 1200 1500 1800 Years CE European summer temperatures 0.5 SAT Europe JJA -0.5 CESM LUT16 -1.5300 600 900 1200 1500 1800 Years CE

- offer:
- new perspectives for combined studies on models/ reconstructions (PAGES2K)
- new insights in origins and effects of 6th century cool phase (aka "Late Antique Little Ice Age")





¹Max Planck Institute for Meteorology, Hamburg, Germany, ²University of Colorado, Boulder, Co, USA ³GEOMAR Helmholtz Centre for Ocean Research, Kiel, Germany, ⁴Helmholtz Zentrum Geesthacht, Germany *Zhong, Y. et al., GRL, 2018 Bader, J., et al. under review

Years CE



Max-Planck-Institut für Meteorologie

Years CE

Ocean heat uptake in the UKESM1 CMIP6 simulations of the historical climate



Till Kuhlbrodt, Colin Jones, Lee de Mora, Julien Palmieri, Andrew Yool

- First analysis of ocean heat content (OHC) anomalies in the CMIP6 historical simulations with the UK Earth System Model UKESM1.
- The EN4.2.1 observational data set shows a ~150 ZJ OHC increase from 1993 to 2015, between 0-2000 m. In the UKESM1 historical simulations the ocean heat uptake is similar (lower panel).
- Both observational data sets show an OHC increase of about ~50 ZJ from 1971 to 1993, between 0-700 m. The UKESM1 historical simulations do not show this (upper panel).
- Reasons for the absence of ocean heat uptake in the simulated 1970s and 1980s are under investigation.





Met Office

Compatible fossil fuel emissions in three CMIP6 Models Spencer Liddicoat, Chris Jones, Andy Wile Models



CMIP6 Historical simulation: results from 3 Earth System Models: UKESM1, IPSL-CM6A-LR, CESM2 • Driven by prescribed historical atmospheric CO₂ • Models simulate carbon uptake by land and by the oceans



• Therefore we can calculate how much fossil fuel CO₂ can be emitted to be consistent with the concentration pathway driving the model.

- These compare well with the observed CO₂ Fossil Fuel emissions.
- This gives us confidence in the ESMs' future carbon budgets.

www.metoffice.gov.uk



Stratosphere-Troposphere Circulation Changes

Manzini, E. (Max Planck Institute for Meteorology, Hamburg, Germany) A. Yu. Karpechko (Finnish Meteorological Institute, Helsinki, Finland)

Motivation: analyze *DynVarMIP/CMIP6* and *single-model large ensembles* to assess, understand and better quantify previously found dynamical uncertainties and their links, with a focus on the stratosphere-surface climate links. (Manzini et al 2014; Simpson et al 2018)

• CMIP models and MPIGE ranges



• Inter-comparison of single-model large ensembles (CMIP6/IPSL and MPIGE)







Present and future seasonal land snow cover simulated by CMIP coupled climate models

Ménégoz M.¹, Krinner G.¹, Brutel-Vuilmet C.¹, Santolaria-Otín M.¹, Derksen C.², Mudryk L.² 1 Institut des Géosciences de l'Environnement (France); 2 Environment and Climate Change Canada



Motivations:

- 45 million km² of the Northern Hemisphere snow-covered in winter.
- Snow albedo feedback, atmospheric circulation, carbon storage in permafrost.



The IPSL-CM6 31 member historical experiment shows a **pronounced retreat of the snow cover extent** (50% climatological level in blue)



Monthly snow cover
decrease as a function of
global temperature
(historical+projection, left)

Internal variability:

31-member distribution of20 year trends in thehistorical experiment (right)

1994-2014 trend of snow cover in NH (%)



Attribution of the observed intensification of extreme precipitation over dry and wet regions

Seungmok Paik and Seung-Ki Min

Pohang University of Science and Technology (POSTECH)





CONTRIBUTION OF THE GREENLAND ICE SHEET TO EUSTATIC SEA LEVEL RISE: PROJECTIONS WITH CMIP6 CESM2.1 – CISM2.1

L. Muntjewerf¹, W.H. Lipscomb², W.J. Sacks², M. Löfverstrom³, J.G. Fyke^{4,5}, R. Sellevold¹, C. Ernani da Silva¹, S.L. Bradley¹, M. Petrini¹, M.Vizcaino¹



Table 1: Greenland cumulative contribution to eustatic sea level rise (mm) for the historical simulation (1850-2014), the SSP1-2.6 scenario (2015-2100), and the SSP5-8.5 scenario (2015-2100). Mass balance (Gt/yr) and





¹ Department of Geoscience and Remote Sensing, Technical University Delft, Delft, The Netherlands. ² Climate and Global Dynamics Laboratory, NCAR, Boulder, USA. ³ Department of Geosciences, University of Arizona, Tucson, USA. ⁴ Associated Engineering Group, Ltd., Edmonton, Alberta, Canada. ⁵ Los Alamos National Laboratory, Los Alamos, NM, USA.

l.muntjewerf@tudelft.nl

Seasonal amplification, phase shift, & uncertainties for ocean acidity during the 21st century *(poster 6-P22)*

J.C Orr (LSCE/IPSL) & L. Kwiatkowski (LMD/IPSL), France





Kwiatkowski & Orr (2018)



Regional analysis of present day marine productivity in UKESM1

J. Palmiéri, A. Yool, E.E. Popova, UKESM1 core group (National Oceanography Centre, Southampton, United Kingdom)



Evaluate UKESM1 bioregions and associated biogeochemistry









160E

120E

- IPWP expansion dominant in Indian Ocean during SON/DJF

7986 १९२०पार्थ

- Robustly attributed to anthropogenic forcing
- Similar results between CMIP5 and CMIP6 models

80W

120W

CMIP5

80E



160W





40E

6-P25: Detecting changes in North Atlantic variability under global warming

90N

60N

30N

30S

60S

90S

-0.3

-0.24 -0.18 -0.12

0.6

0.4

0.2

0.0

-0.2

-0.4

-0.6

0.2

-0.2

-0.4

-0.6

Putrasahan, D. A., Jungclaus, J. H., von Storch, J.-S., Ghosh, R.

Max Planck Institute for Meteorology, Hamburg, Germany

The Max Planck Institute - Grand Ensemble (MPI-GE) gives us the opportunity to assess changes in internal variability in a transient climate. This is done by extending 2 classical techniques (simple EOFs and cross spectral analysis) to:

- EOFs in ensemble space to primarily detect changes in 1) spatial patterns of dominant modes (e.g. North Atlantic SST)
- 2) squared coherence between a climate index (e.g. NAO) and a climate variable (e.g. DJF surface temperature) all over the world to evaluate changes in the pattern of their relationship at different timescales











Max-Planck-Institut für Meteorologie



0.06 0.12 0.18 0.24

0.3

0

-0.06



Assessing co-behavior of climate processes over southern Africa using CMIP5 models [S6 P26]

¹Kwesi A. Quagraine*, ¹Bruce Hewitson, ¹Chris Jack, ¹Izidine Pinto, ¹Chris Lennard ¹Piotr Wolski



¹Climate Systems Analysis Group, University of Cape Town, South Africa

(contact: kwesi@csag.uct.ac.za)

Background

- Regional climates → a no. of climate processes operating in multiple spatial and temporal scales.
- Evaluating the regional response to the collective co-behavior of these processes is central to understanding the region's climate, more so with regions with no dominant large-scale driver, this is important.
- Co-behavior is a concept used here and is interpreted as an interaction between at least two or more climate features leading to their influence on the weather and climate for any given region.

Data

- Observational datasets (Precipitation → CHIRPS and temperature → CRU)
- Climate indices for Antarctic Oscillation (AAO), El Niño Southern Oscillation (ENSO) and Inter-Tropical Convergence Zone (ITCZ)
- 8 CMIP5 GCMs

Methods

Self-Organising Map (SOM), PCA and composite analysis

Summary

- AAO) moderates the regional precipitation and temperature response to EL Niño when co-behaving
- CMIP5 models largely agree with the sign of change of identified cobehavior modes in observational datasets.





Attribution of record-breaking hot summer over Northeast Asia in July-August 2018: the contribution of circulation

Liwen Ren, Tianjun Zhou

Institute of Atmospheric Physics, Chinese Academy of Sciences, China





Tracking the impact of climate model complexity in future climate projections using CNRM-ESM2-1 and CNRM-CM6-1



Roland Séférian, Pierre Nabat, Martine Michou, David Saint-Martin, Aurore Voldoire, Jeanne Colin, Bertrand Decharme, Christine Delire, Sarah Berthet and the CNRM-CERFACS Modelling group



The lifetime of fossil-fuel derived carbon

Atmospheric lifetimes from impulse-response experiments



Assumptions:

- Linearity
- Steady-state
- Non-transient simulations

New approach based on model reconstructions from numerical output



Objectives

- Quantify time to remove fossil-fuel carbon
- Compare among models and scenarios
- Quantify forward and transit times



1 / 1

Session 6, Poster 29



What's up with what's going down? **Reading** Trends in primary and export production



Andrew Yool, Julien Palmiéri, Katya Popova, Lee de Mora, Alistair Sellar, Colin Jones, the UKESM1 Core Group, Roland Séférian, Sarah Berthet, Yohei Takano

- Climate change is coming (it's here); the biological pump is threatened
- Carbon flux plays a role in both ocean carbon storage and in the supply of food to deep seafloor communities



