

# **Analysis of atmospheric circulation from climate model big data - Current approaches and future challenges**

**Dörthe Handorf, Klaus Dethloff, Ralf Jaiser,  
Annette Rinke**

**Section Physics of the Atmosphere**

**Research Department Potsdam, Germany**

**Alfred Wegener Institute**

**Helmholtz Center for Polar and Marine Research**

**AWI Data Science Symposium**

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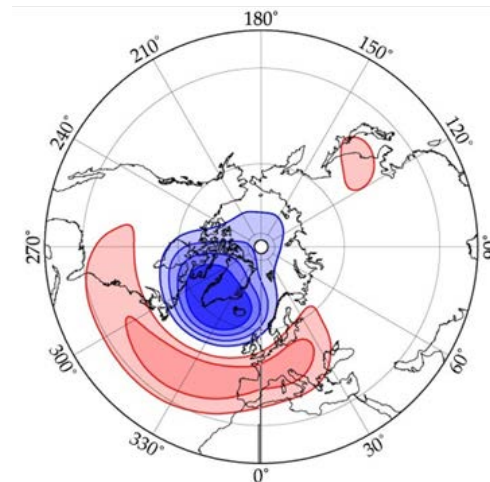
# Patterns of atmospheric variability – Atmospheric teleconnections



- preferred patterns of low-frequency variability (subseasonal to decadal timescales)
- refer to recurring and persistent, large-scale patterns of pressure and circulation anomalies that spans vast geographical areas
- are localised in definite regions (hemispheric-scale, basin-wide, continental)
- are a naturally occurring aspect of our chaotic atmospheric system
- can arise primarily due to the internal atmospheric dynamics, but can be impacted by external forcings
- reflect large-scale changes in the atmospheric wave and jet stream patterns
- influence temperature, rainfall, storm tracks over vast areas

## → Example: North Atlantic Oscillation (NAO)

Most important pattern over North-Atlantic-European Region  
Dipole structure over North Atlantic  
associated with changes in location/intensity of North Atlantic jet stream and storm tracks

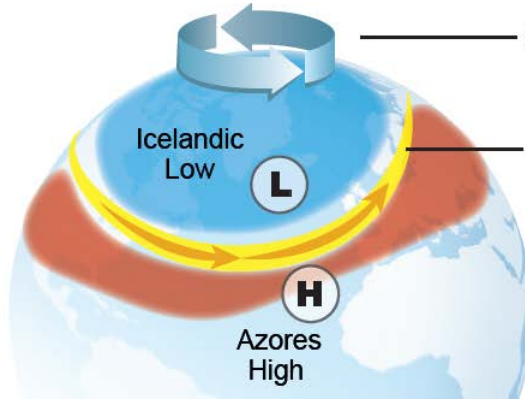


# Patterns of atmospheric variability - N-Atl./Europe

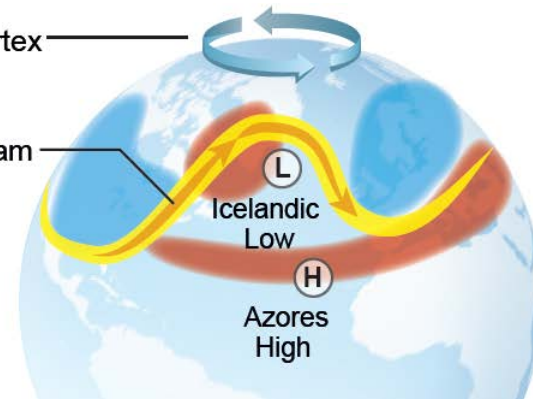
## The North Atlantic Oscillation (NAO)

### Two states of atmospheric circulation over the North Atlantic-European sector

**Zonal jet stream**  
Small-amplitude planetary waves



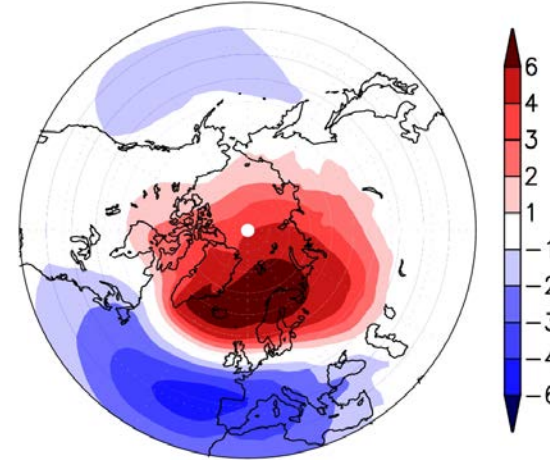
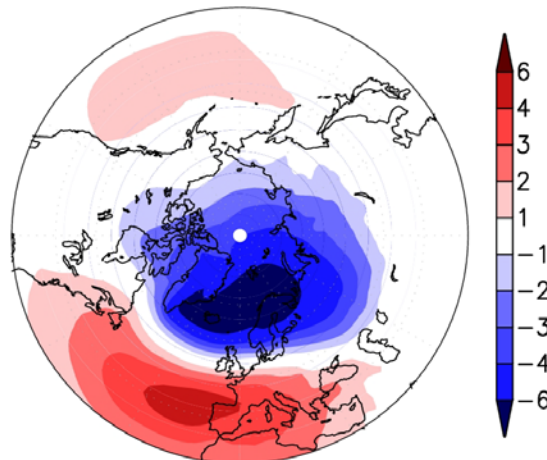
Polar Vortex  
Jet Stream



**Meandering jet stream**  
Large-amplitude planetary waves

Corresponding patterns of sea-level pressure anomalies (deviation from mean pressure distribution)

North-Atlantic Oscillation in positive phase (NAO+)



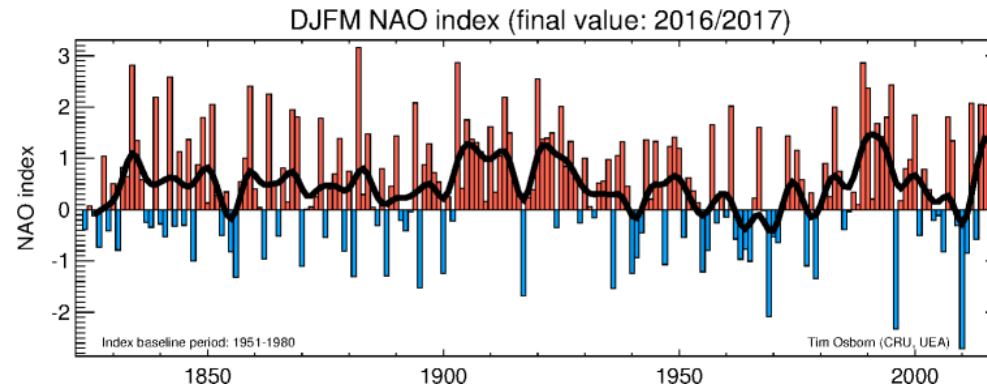
North-Atlantic Oscillation in negative phase (NAO-)

# Patterns of atmospheric variability - N-Atl./Europe

## The North Atlantic Oscillation (NAO)

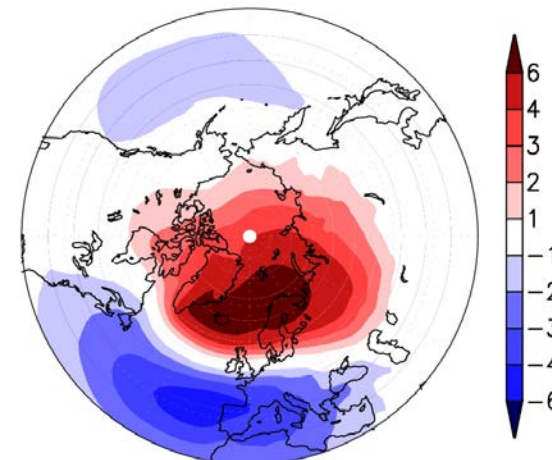
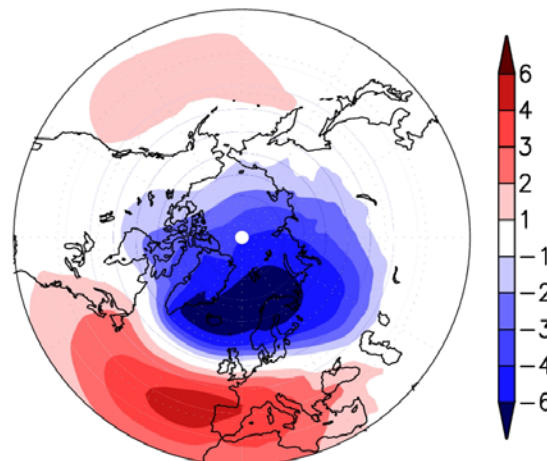
### Two states of atmospheric circulation over the North Atlantic-European sector

Interannual to decadal variability



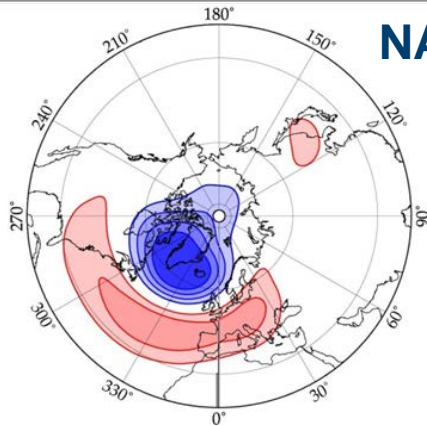
Corresponding patterns of sea-level pressure anomalies (deviation from mean pressure distribution)

North-Atlantic Oscillation in positive phase (NAO+)



North-Atlantic Oscillation in negative phase (NAO-)

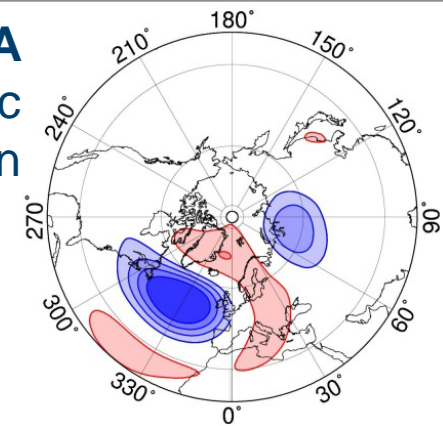
# Patterns of atmospheric variability – Research Questions



## NAO

- associated with changes in the North Atlantic jet stream and storm tracks

## EA East-Atlantic pattern



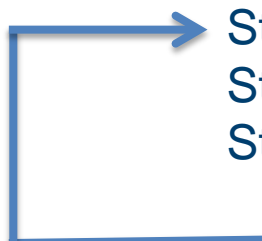
## Research Questions

- Understand past, recent future changes in the spatial/temporal structure of teleconnection patterns
- Internal dynamics versus external forcing
- Potential predictability of teleconnections

## Global, gridded data sets

- Reanalysis data sets
- Climate model simulations

## Hypothesis-driven Approach

- 
- A flowchart illustrating the hypothesis-driven approach. It starts with a large blue arrow pointing right towards the first step. A vertical line descends from the end of this arrow, and a horizontal line extends from the bottom of this vertical line back to the start of the first arrow, forming a U-shape. This indicates a feedback loop from the final step back to the first step.
- Step 1: Analysis of changes – Evaluation of climate models
  - Step 2: Development of hypothesis
  - Step 3: Provision of evidence
    - New model experiments
    - New analysis

# Global gridded climate data sets: Reanalysis and Earth system models



## Data amount of Reanalysis:

→ Example: ECMWF, global

ERA40	1957/09 to 2002/08			
	Sub-daily, Monthly	2.5°x2.5° / 1.125°x1.125°	60 levels	0.1 hPA top
ERA-Interim	1979/01 to present			
	Sub-daily, Daily, Monthly	0.75°x0.75°	60 levels	0.1 hPA top
ERA5	1979/01 to present			
	hourly, Sub-daily, Daily, Monthly	0.28°x0.28°	137 lev	0.01 hPA top

Expected: 5 Petabytes for ERA5

# Global gridded climate data sets: Reanalysis and Earth system models

## Data amount Reanalysis:

Example: ECMWF, global → ERA5 → 5 Petabytes

## Data amount CMIPs Coupled model intercomparison projects

- **CMIP3** (for IPCC AR4 2007): 17 institutes (groups) and 25 models → 40 TB
  - total years simulated: 70000
  - individual models simulated on average 2800yrs
- **CMIP5** (for IPCC AR5 2013): 26 institutes (groups) and 60 models → 2 PB
  - total years simulated: 330000
  - individual models simulated on average  $330000/60 = 5500$  years
  
- Extrapolation for **CMIP6** data federation:
  - CMIP6 has a more complex experiment structure than CMIP5
  - 32 institutes (groups) and many model versions
  - more models with higher resolution models
  - 21 MIPs, many experiments, larger ensembles
  - Expectations:
    - Volume: 150 PB
    - Number of files: 280 Mio Files

→ **Climate big data?!**

### ➤ Empirical Orthogonal Function Analysis (EOF)

- reduce the dimensionality of the data
- find the most important patterns explaining the variability
- provide information about spatial structures and temporal scales

➤ Data field represented compactly in terms of EOFs:  $Q'_i(t) = \sum_{j=1}^N \alpha_j(t) e_{ij}$

### ➤ Principal components $\alpha_j(t)$

- represent projections of data onto the  $j$ -th EOF

### ➤ Rotation of EOF produce more localised patterns

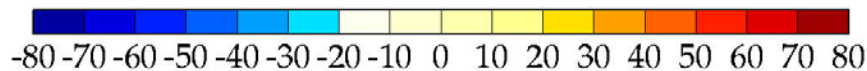
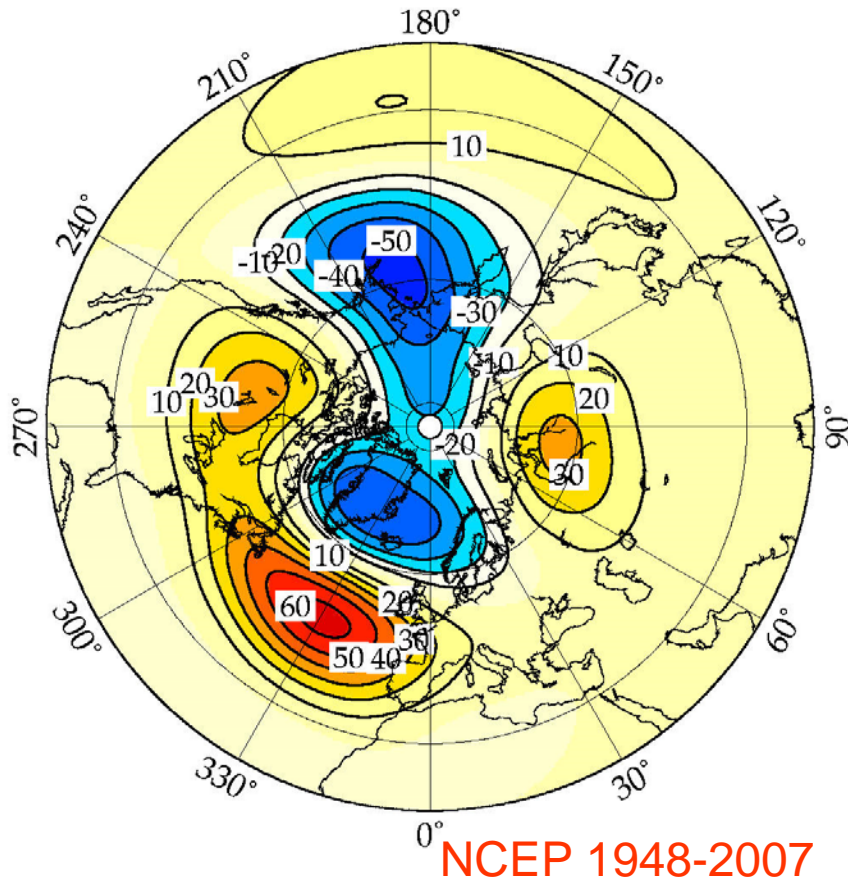
- Rotated EOFs = linear combination of first few EOFs, determined by minimisation of a functional (e.g. spatial variance)



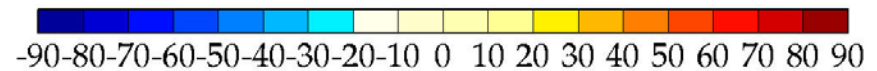
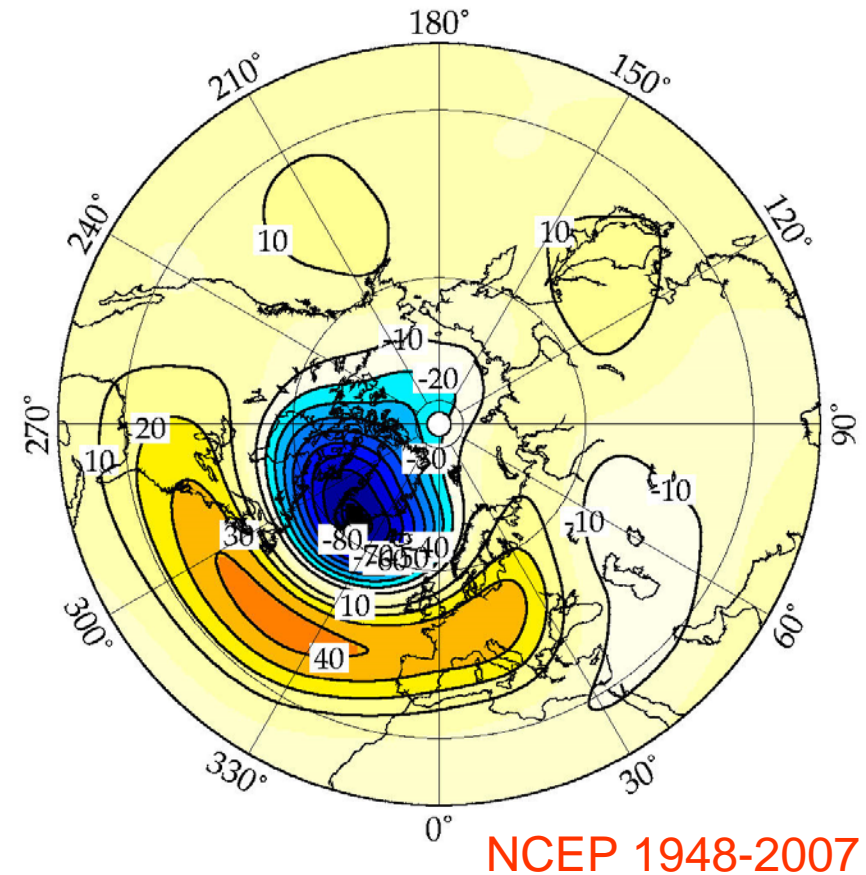
# Analysis of teleconnection patterns

## Empirical Orthogonal Function Analysis

EOF2 12.4% dominated by NAO



ROT EOF1 13.4% NAO



# Analysis of teleconnection patterns in Reanalysis data and Evaluation of Climate models



- Analyses of monthly mean data of midtropospheric circulation  
→ 500hPa geopotential height fields
- Analyses of dynamically active season of Northern Hemisphere (NH)  
→ December, January, February data (DJF)
- Fields from 20°-90° N with removed seasonal cycle
- Evaluation with NCEP/NCAR and ERA40 Reanalyses

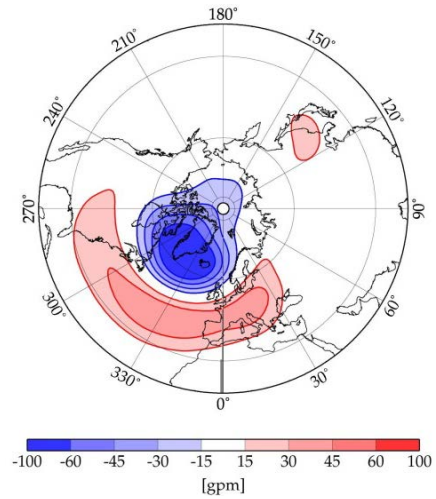
## Analysed Experiments from CMIP3/CMIP5

<b>CMIP3</b> 20 <sup>th</sup> cen. simulation 1870-1999 Analyses of years 1958-1999	forced by observed atmospheric composition changes (anthropogenic & natural sources)	23 models
<b>CMIP5</b> historical simulations 1850-2005 Analyses of years 1958-1999	forced by observed atmospheric composition changes (anthropogenic & natural sources) time-evolving land cover	46 models more comprehensive generally with higher spatial resolution

### Technological approach

- Download of data from data centers
- Data analysis locally
- Software packages (MATLAB, R)
- Own software (FORTRAN, R) → use of libraries (NAG)

### Reanalysis ERA40

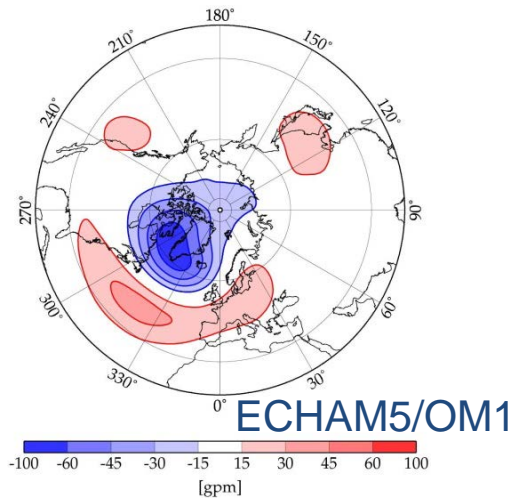
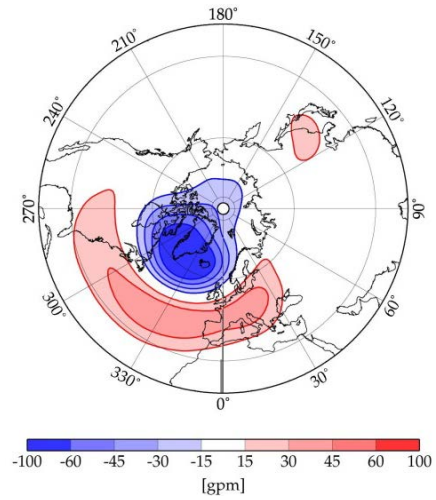


# Teleconnections - Evaluation of spatial structure

## CMIP3 ensemble - Period 1958-1999 - NAO

### Reanalysis ERA40

### Single Model



# Teleconnections - Evaluation of spatial structure

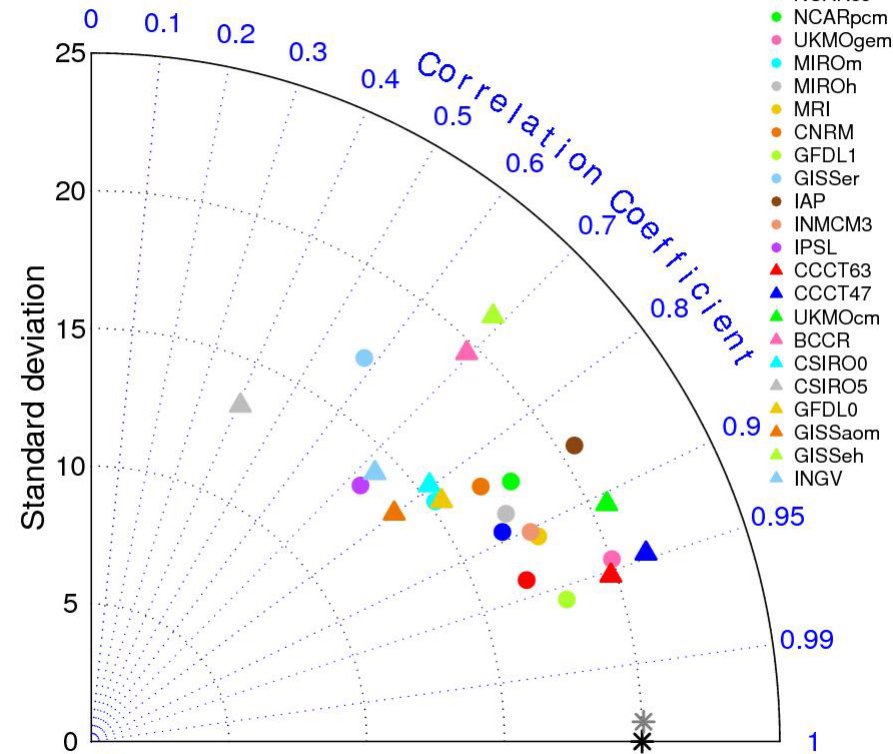
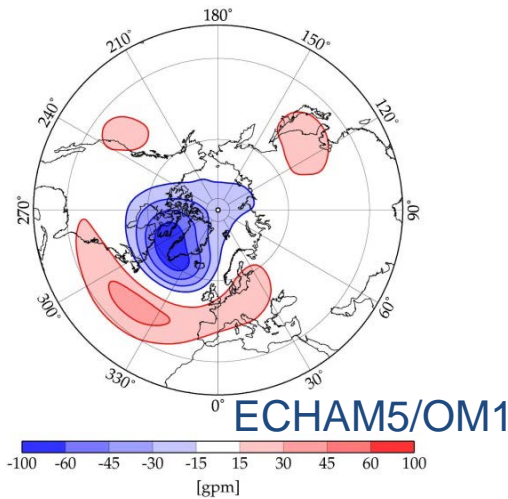
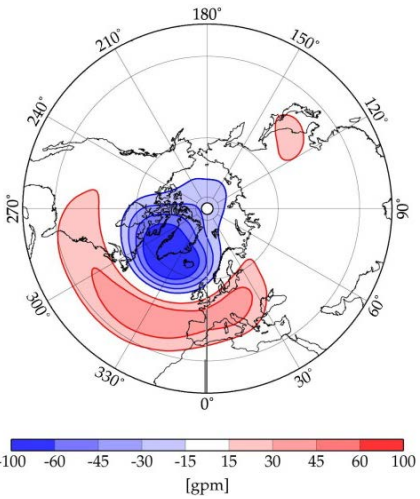
## CMIP3 ensemble - Period 1958-1999 - NAO

### Reanalysis ERA40

### Single Model

### All Models

- \* ERA40
- \* NCEP
- MPI
- NCARcc
- NCARpcm
- UKMOgem
- MIROm
- MIROh
- MRI
- CNRM
- GFDL1
- GISSer
- IAP
- INMCM3
- IPSL
- ▲ CCCT63
- ▲ CCCT47
- ▲ UKMOcm
- ▲ BCCR
- ▲ CSIRO0
- ▲ CSIRO5
- ▲ GFDL0
- ▲ GISSaom
- ▲ GISSeh
- ▲ INGV



### Taylor diagrams (Taylor, 2001)

- Quantify similarity between different patterns
- Compact summary of pattern statistics in terms of pattern correlation, root-mean-square difference and ratio of variances.

# Teleconnections - Evaluation of spatial structure

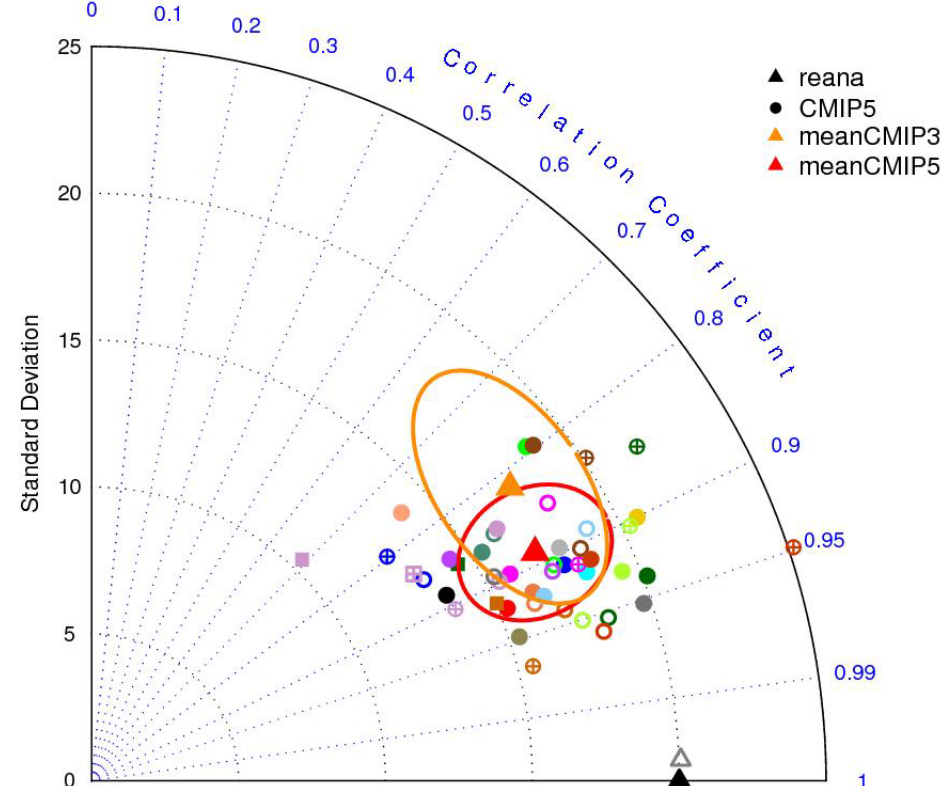
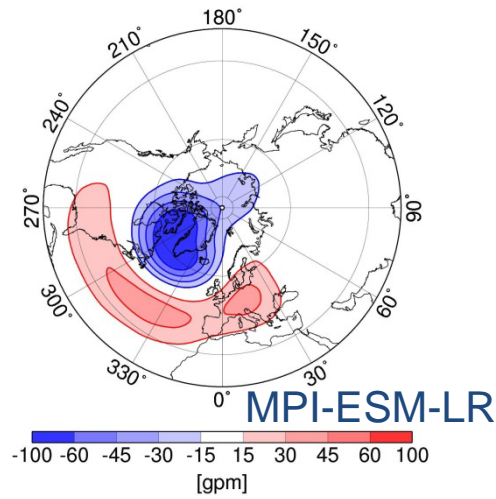
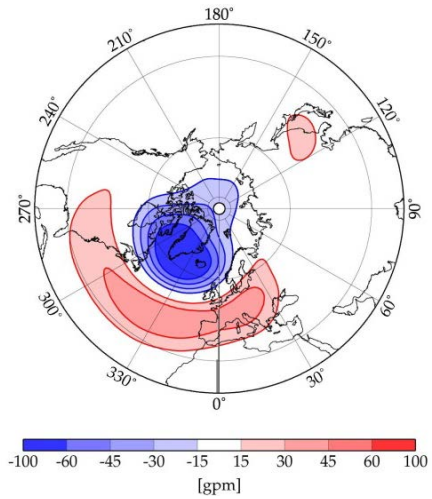
## CMIP5 ensemble - Period 1958-1999 - NAO



### Reanalysis ERA40

### Single Model

### All Models



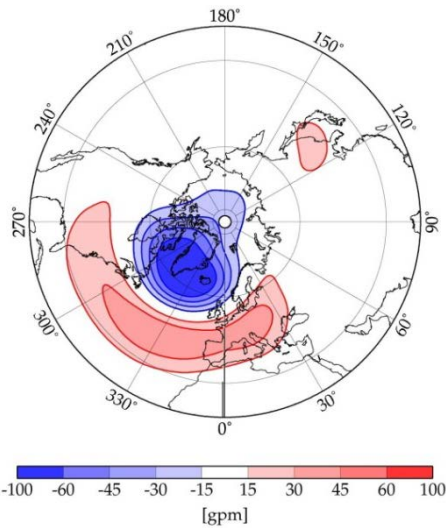
Spread in the skill of simulating spatial patterns over the CMIP5 ensemble  
Improvement

- |             |            |            |            |             |            |
|-------------|------------|------------|------------|-------------|------------|
| ▲ ERA40     | ○ GFDLcm3  | ○ HADgemAO | ● IPSL5AMR | ○ MIROesm   | ○ NorESMme |
| △ NCEP      | ● GFDLesmG | ● HADgemCC | ● MPL_LR   | ● MIROchem  | ● BCC      |
| ● CANesm    | ■ GFDLesmM | ■ HADgemES | ● MPL_MR   | ● NCAR      | ● BCCm     |
| ● CNRM      | ● GISSeh   | ● IAPg     | ● MPL_P    | ● CESM_BGC  | ● BNU      |
| ● CSIROacc0 | ● GISSehCC | ● IAPs     | ● MRI      | ● CESM_CAM  | ● CMC      |
| ● CSIROacc3 | ● GISSer   | ● INM      | ● MRlesm   | ● CESM_FAST | ● CMCS     |
| ● CSIRO36   | ● GISSerCC | ● IPSL5ALR | ● MIROh    | ● CESM_WAC  | ● CMC_ESM  |
| ● GFDLcm2   | ● HADcm    | ● IPSL5BLR | ● MIRO5    | ● NorESMm   | ● FIO      |

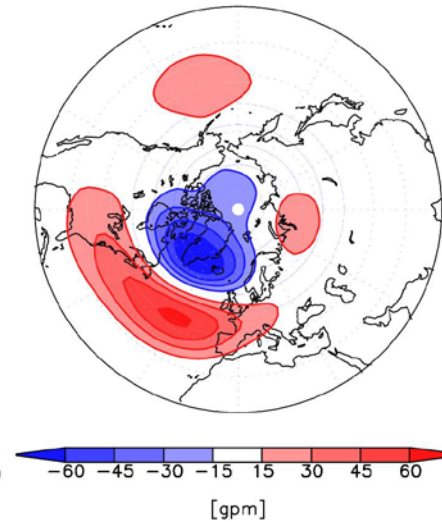
# Dynamical reasons for limited skill of the CMIP3/5 ensemble in reproducing teleconnections

- Hypothesis 1: Deficiencies in atmospheric internal dynamics
- Teleconnections are related to variability of zonal wind **for the reanalyses**  
(gph and u fields are dynamically related, e.g., Athanasiadis et al., 2010; Li and Wettstein, 2012)

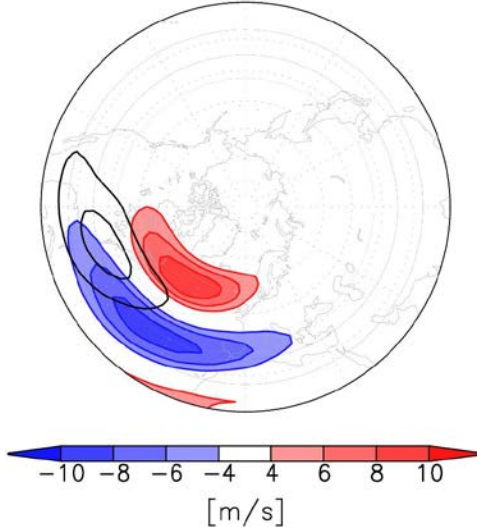
**NAO** pattern  
@500hPa



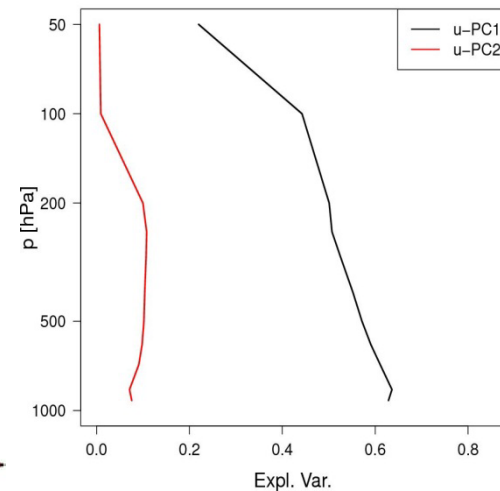
**Regression** onto Atlantic  
u-PC1 @250hPa for GPH500



**EOF1** of Atlantic zonal  
wind @250hPa



Profile of explained  
variance



- NAO** closely related to EOF1 of Atlantic zonal wind (Position of eddy-driven jet)
- EA** closely related to EOF2 of Atlantic zonal wind (Intensity of eddy-driven jet)

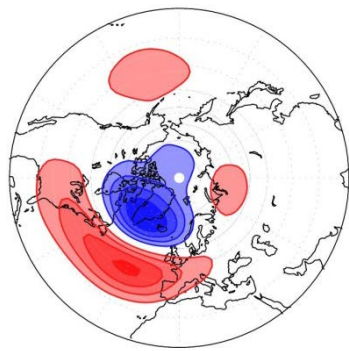
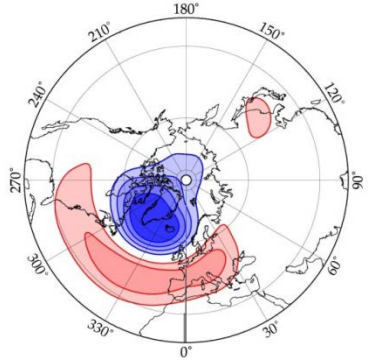
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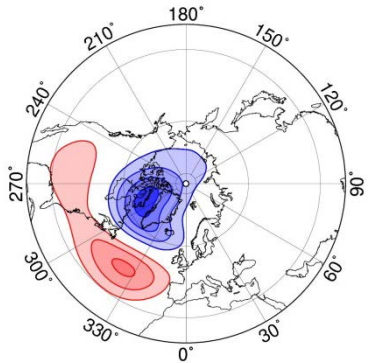
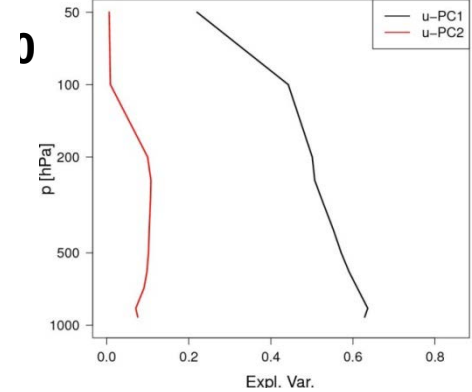
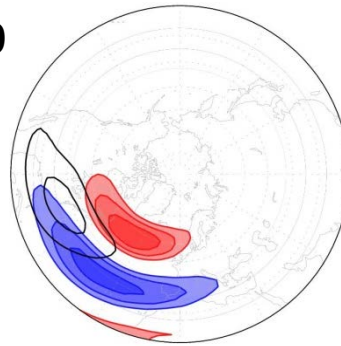
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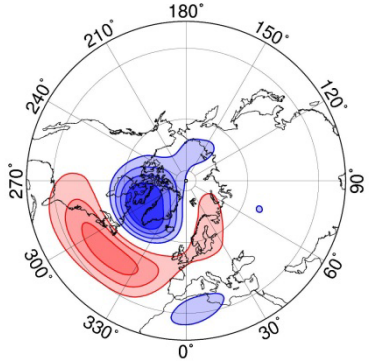
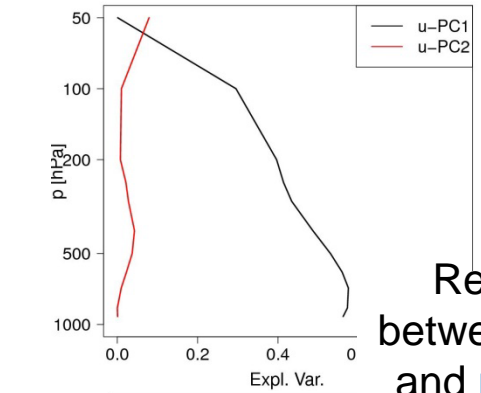
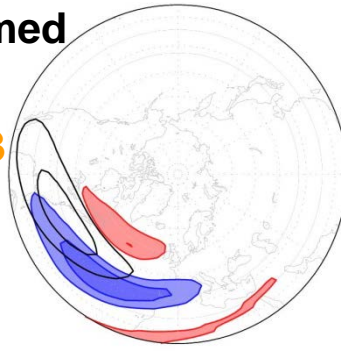
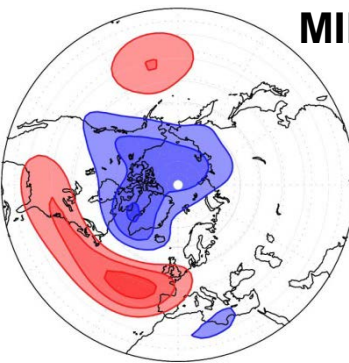
Profile of explained  
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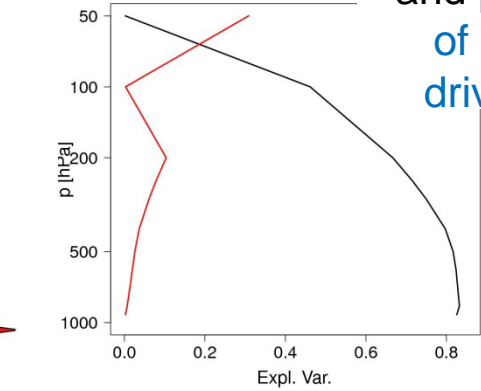
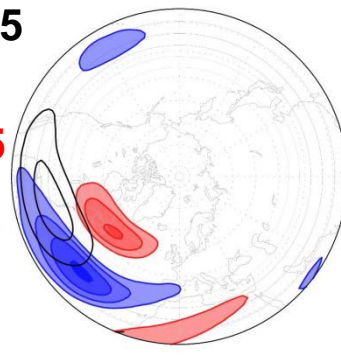
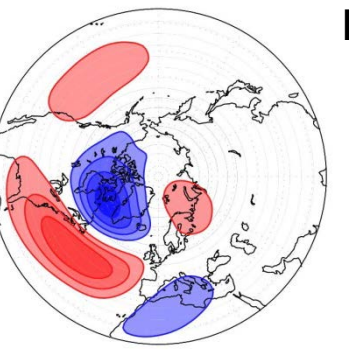
**ERA40**



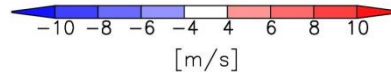
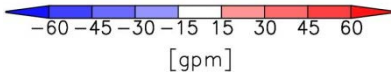
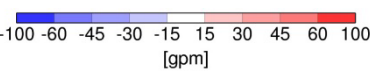
**MIROC\_med**  
from  
**CMIP3**



**MIROC5**  
from  
**CMIP5**



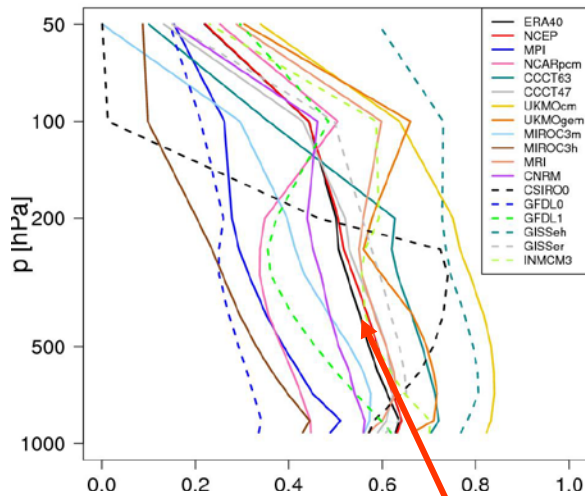
Relation  
between **NAO**  
and **position**  
of eddy-driven  
jet



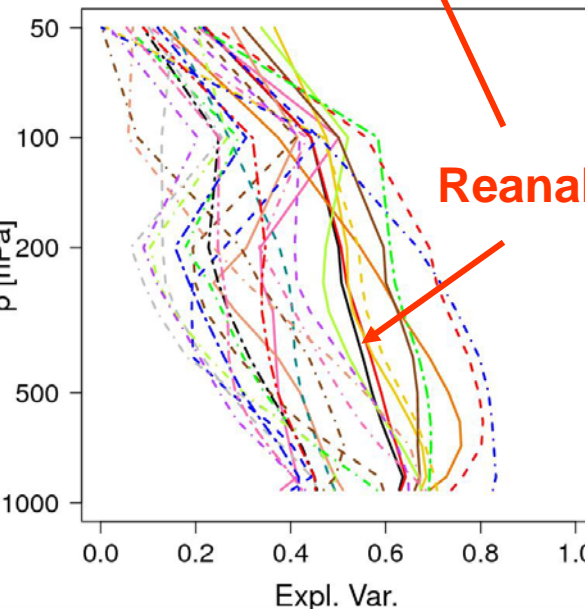


# Dynamical reasons for limited skill of the CMIP3/5 ensemble in reproducing teleconnections

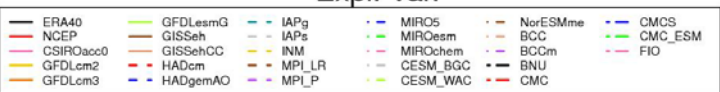
## Explained variance with NAO



Structure of the relation between teleconnections and zonal wind variability captured by **some (not all) models** of the CMIP3 and CMIP5 ensemble (large spread)

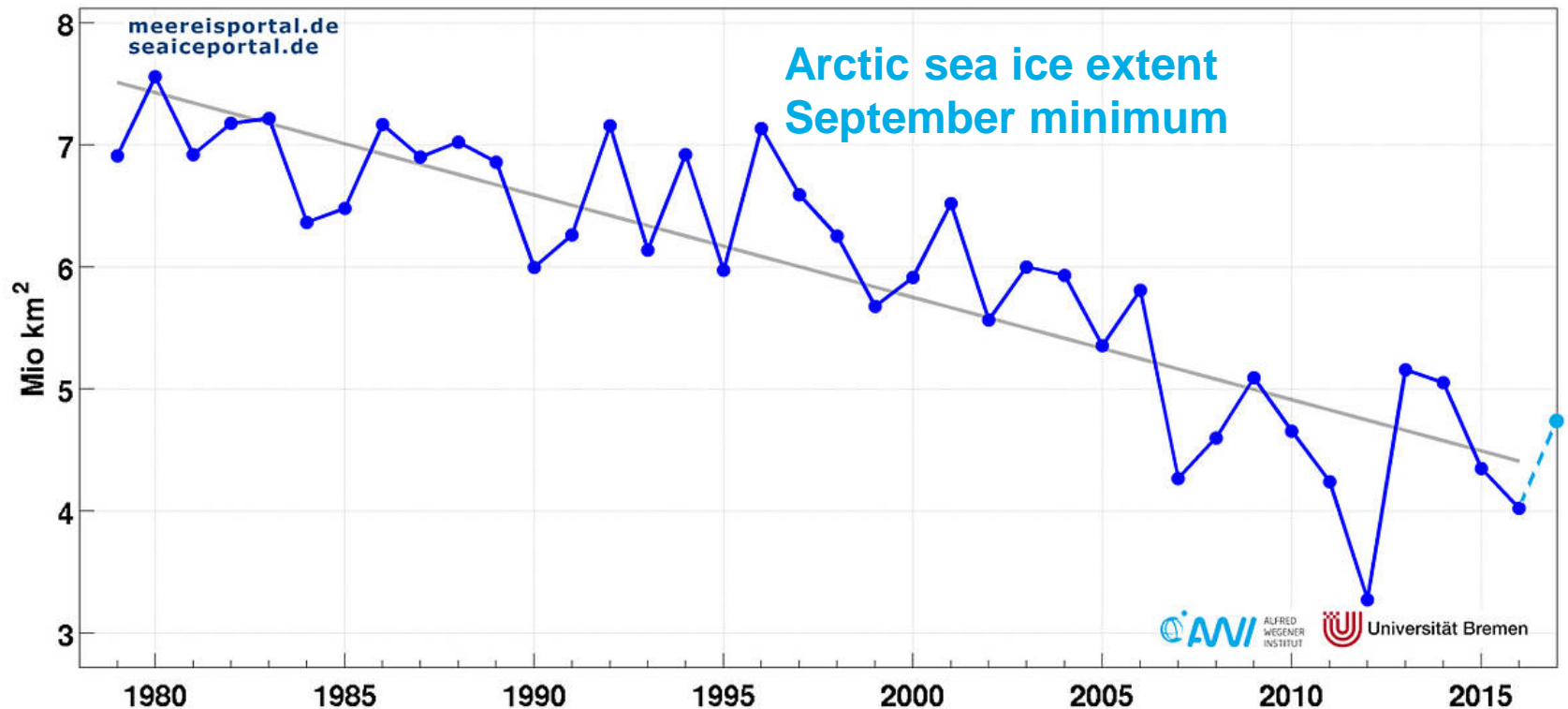


The quality of the simulated teleconnection pattern is largely determined by the quality of the simulated zonal wind variability pattern



# Impact of external forcings on teleconnections and their reproduction in climate models

- Hypothesis 2: Preferred state of atmospheric variability patterns is influenced by external forcing factors
- Example: Forcing due to **changes in sea-ice**



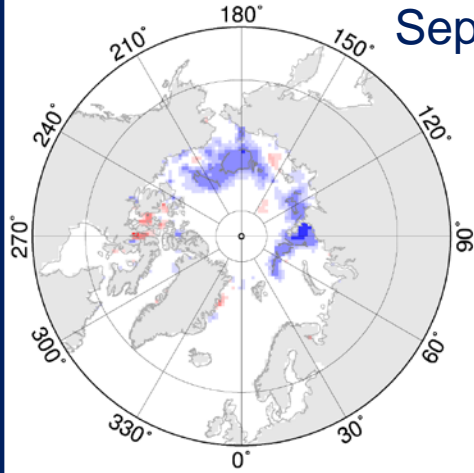
- Hypothesis 2: Preferred state of atmospheric variability patterns over the North-Atlantic-Eurasian region is influenced by Arctic climate changes (e.g. sea-ice changes)

## **Methods: Maximum Covariance Analysis (MCA):**

- Statistical method detecting coupled patterns between pairs of climate fields
- Maximized covariance of time series associated to each pattern
- **Reanalysis data: ERA-Interim**
  - September sea ice concentration 1979-2015
  - Mean sea level pressure and geopotential height fields in Winter (February or DJF, 1979-2015)
- **Climate model data:**
  - AFES (Atmospheric general circulation model For Earth Simulator)
  - Ensemble model simulations, 30 members
  - AMIP-style, 1979-2014

# Impact of sea-ice changes on atmospheric teleconnections – Reanalysis

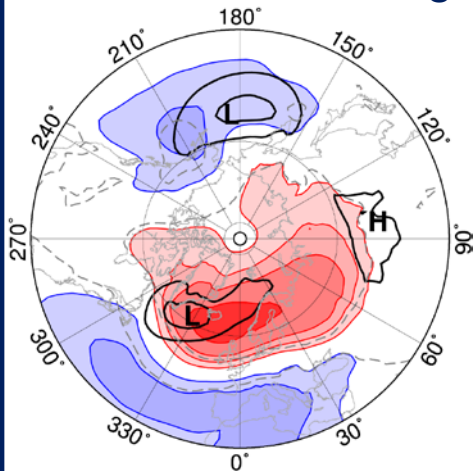
## Sea ice concentration September (HadISST Data)



[-10 -6 -3 -2 -1 1 2 3 6 10] [%]

## Sea level pressure Following February (ERA-Interim)

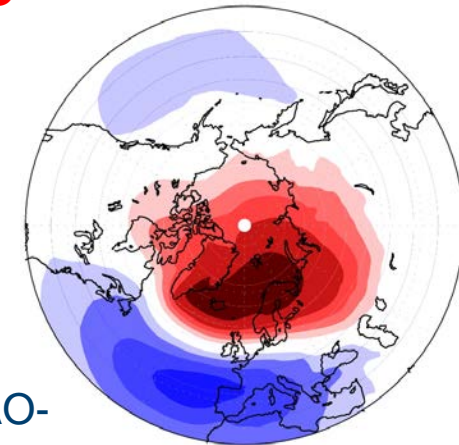
Following February (ERA-Interim)



[-8 -5 -3 -2 -1 1 2 3 5 8] [hPa]

## Planetary-scale response in February Coupled Patterns 1979-2015

- Statistical relation between sea ice retreat and changes of atmospheric circulation patterns
- Changes of centers of action, similarity with pattern of **NAO** in negative phase

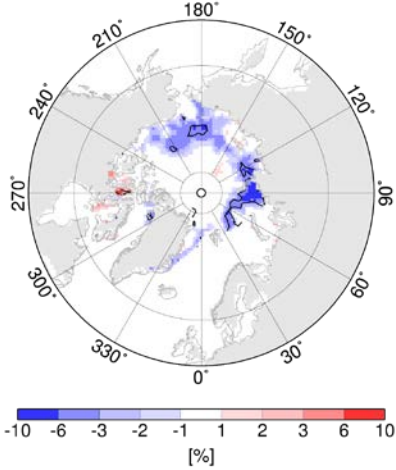


Pattern of NAO-

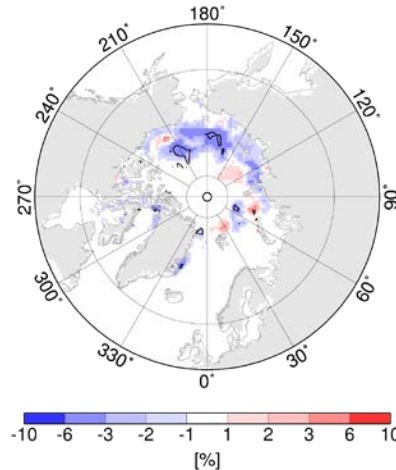
# Impact of sea-ice changes on atmospheric teleconnections – Reanalysis

Sea ice concentration

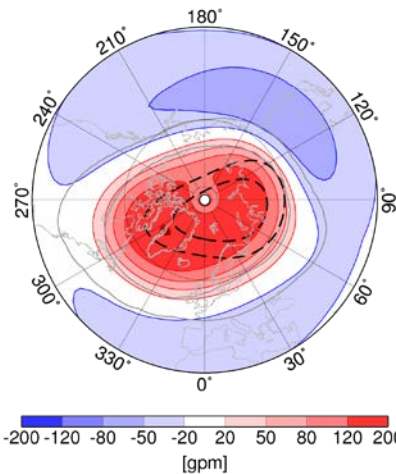
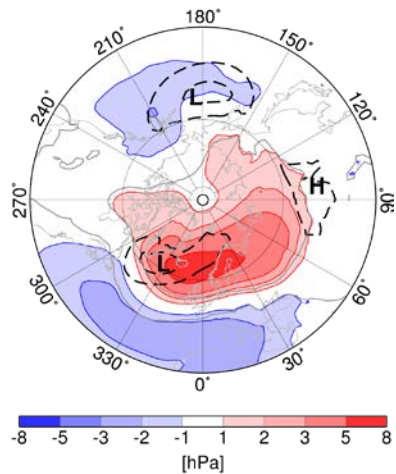
September



September



Atmospheric circulation



February

Sea level pressure

39% explained Covariance

February

GPH 50hPa

61% expl. Covariance

## Planetary-scale response in Feb. Coupled Patterns 1979-2015

- Statistical relation between sea ice retreat and changes of atmospheric circulation patterns
- Changes of centers of action, similarity with pattern of **NAO** in negative phase
- Associated changes in stratosphere → Weaker stratospheric Polar Vortex

# Impact of sea-ice changes on atmospheric teleconnections – Ensemble model simulations

Sep Sea ice concentr.

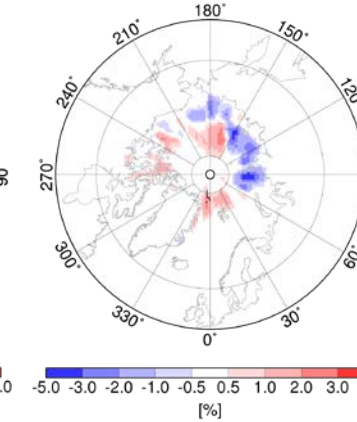
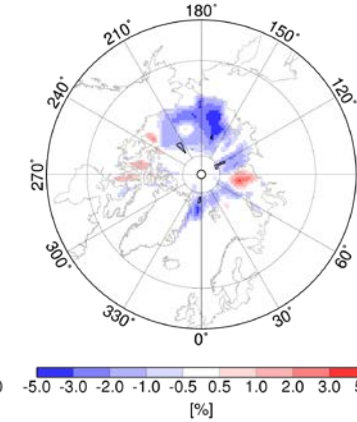
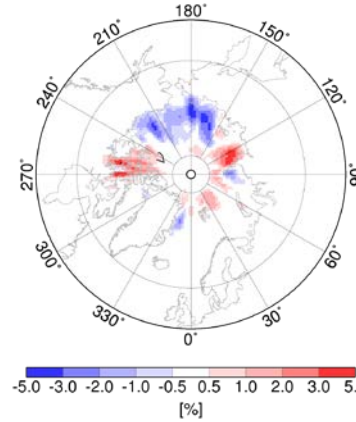
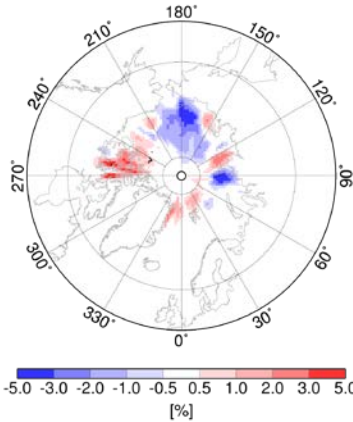
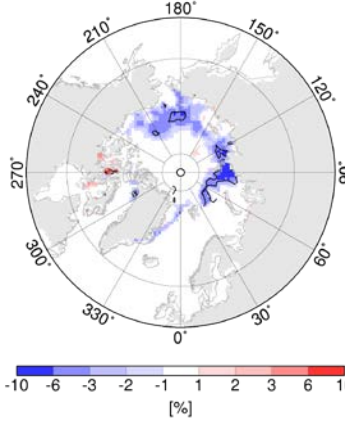
**IERA**  
September

**AFES Run 09**

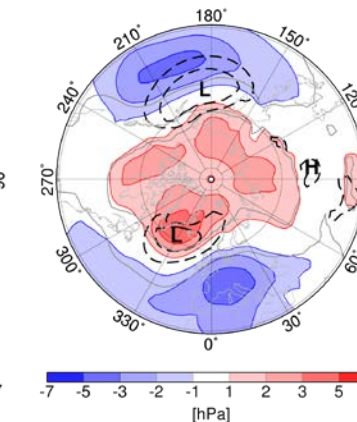
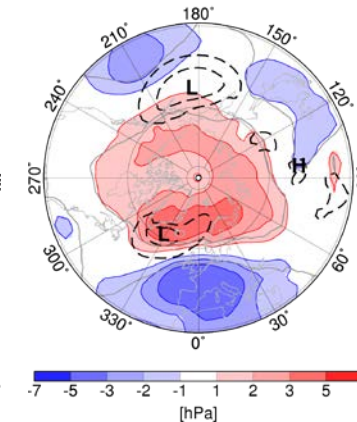
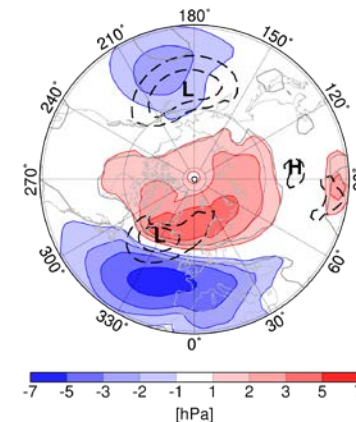
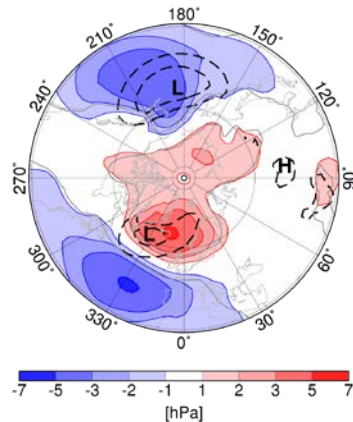
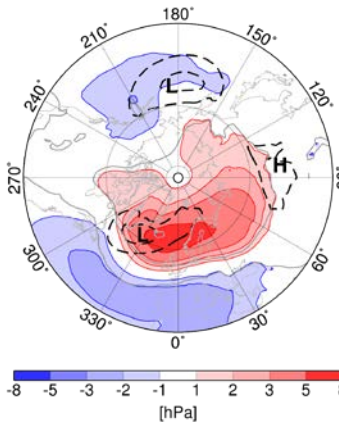
**AFES Run 10**

**AFES Run 17**

**AFES Run 21**



Feb Sea level pressure



39% expl. Covar.

32% expl. Covar.

30% expl. Covar.

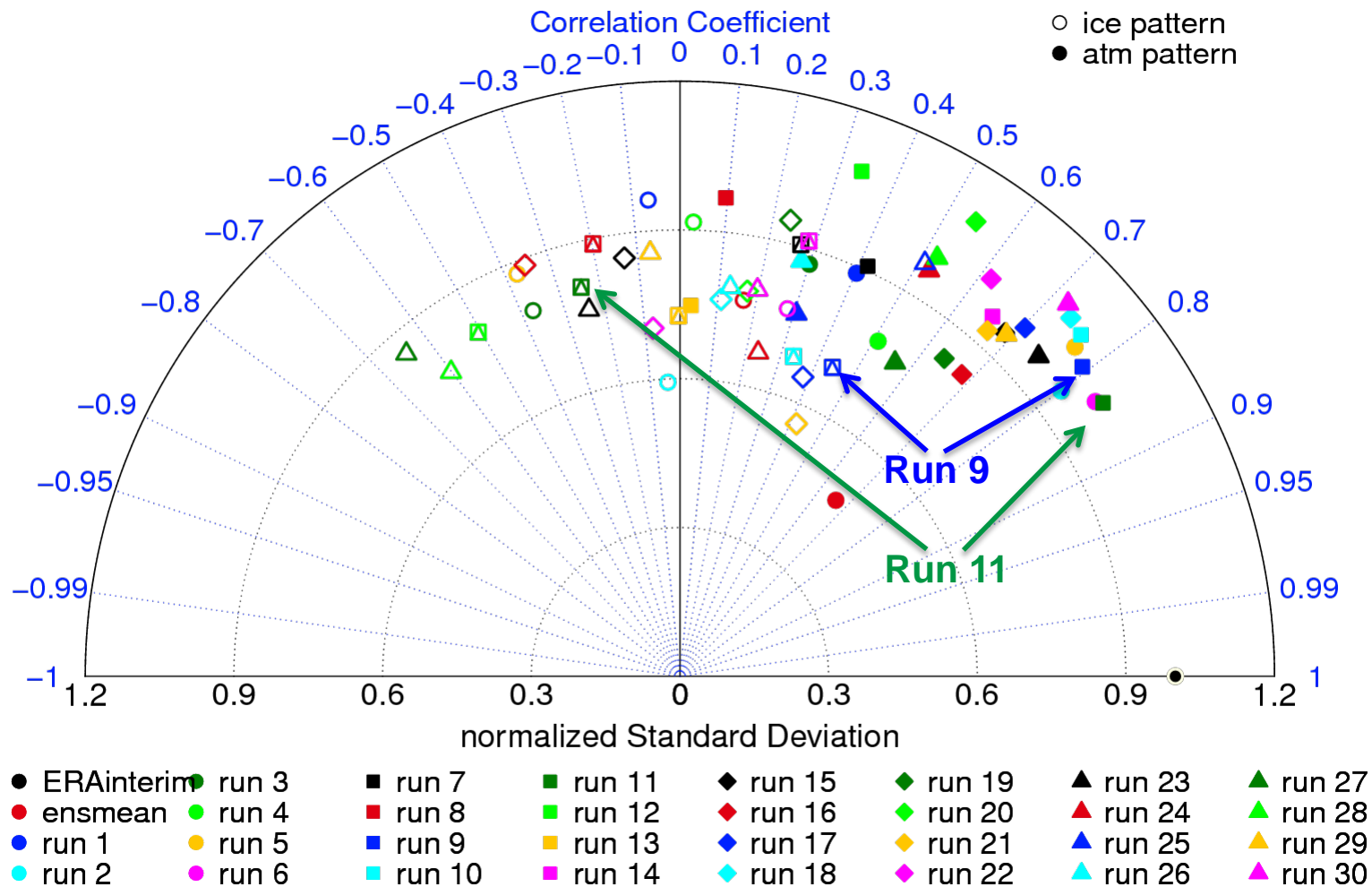
32% expl. Covar.

25% expl. Covar.

- Model: AFES (Atmospheric general circulation model For Earth Simulator)
- Ensemble model simulations, 30 members, AMIP-style

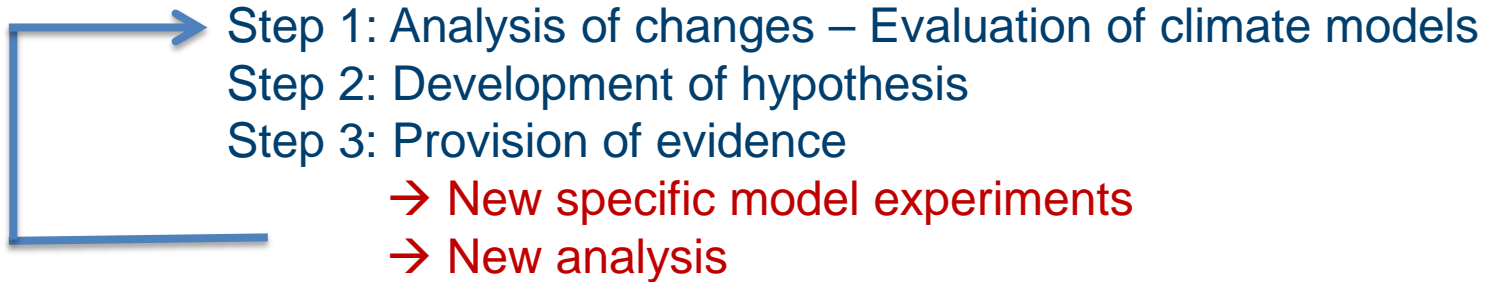
# Impact of sea-ice changes on atmospheric teleconnections – Ensemble model simulations

- Coupled Patterns from ensemble model simulations → Taylorplot
- AFES Ensemble model simulations, 30 members



# Impact of sea-ice changes on atmospheric teleconnections

## Hypothesis-driven Approach



## Set-up of new specific model experiments

- Model: AFES → 2 model simulations, with 60 perpendicular years each
  - CNTL: High ice conditions as observed from 1979 to 1983
  - NICE: Low ice conditions as observed from 2005 to 2009
  - **Only sea ice is different between both runs**

## Maps of sea ice concentration in fall (SON) for low minus high ice conditions





# Impact of sea-ice changes on atmospheric teleconnections

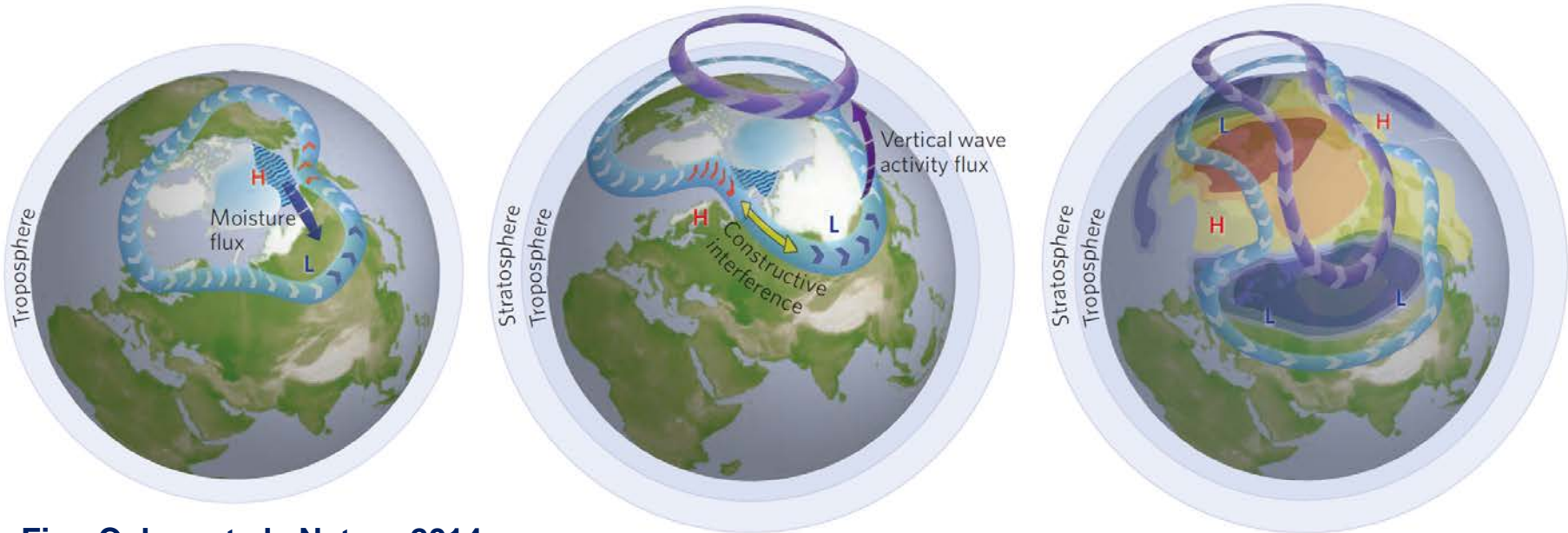


Fig.: Cohen et al., Nature 2014

Sept.

Oct.

Nov.

Dec.

Jan.

Feb.

## Sea ice retreat

- Vertical heat- and moisture fluxes
- Increased baroclinic instability (cyclones)
- Increase in snow cover over Siberia

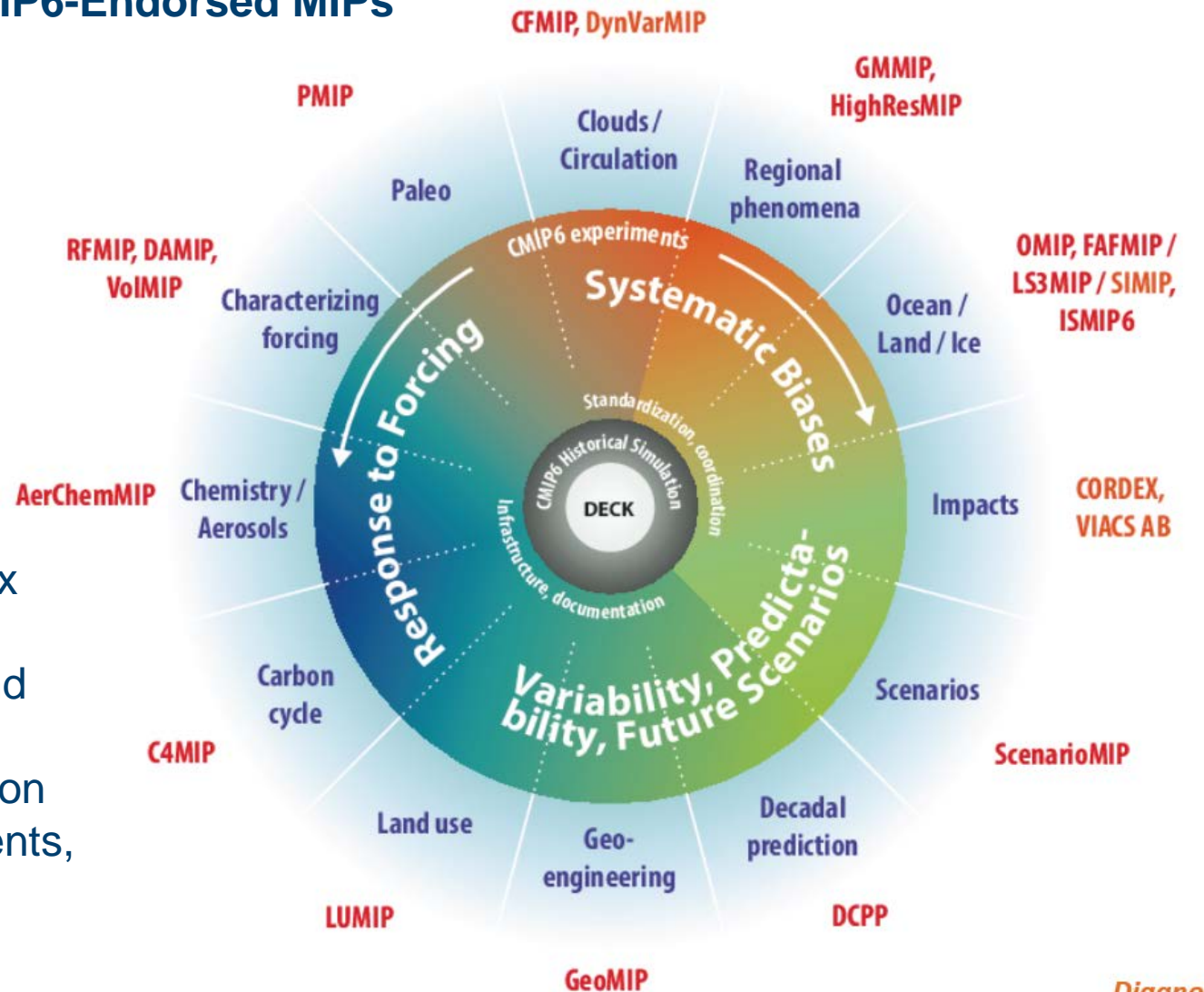
## Forcing of planetary waves

- Interactions between planetary and synoptic waves
- Diabatic forcing due to
  - changes in snow cover
  - ice anomalies in Nov.
- Decreased meridional temperature gradient

## Enhanced planetary waves

- Enhanced vertical wave-propagation up to the stratosphere (EP-fluxes)
- Disturbance of stratospheric polar vortex
- Downward propagating signal
- negative NAO
  - colder European winter

## 21 CMIP6-Endorsed MIPs



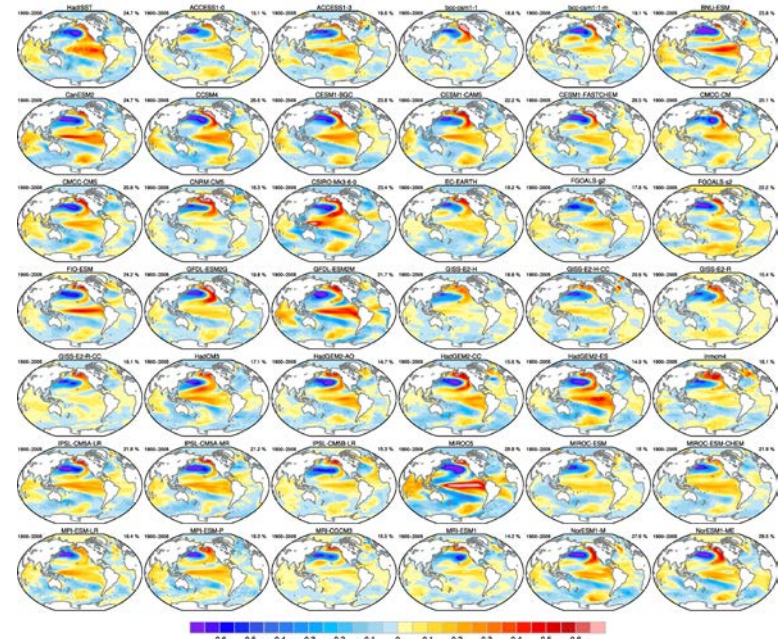
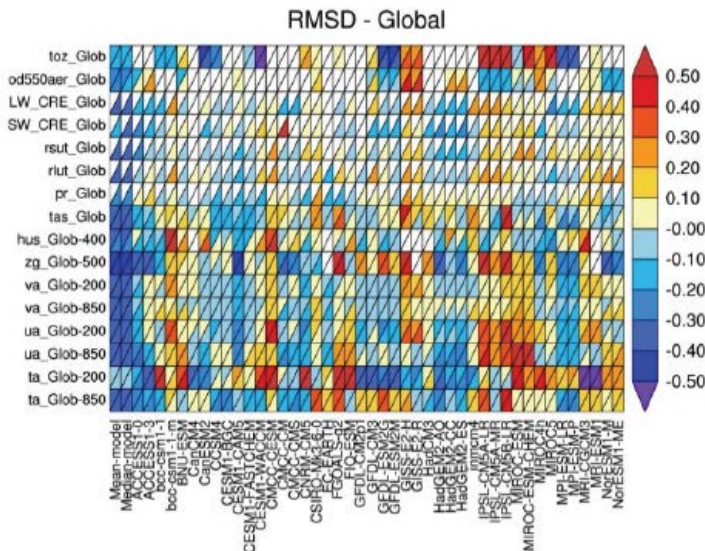
CMIP6 has a more complex experiment structure  
 → 32 institutes (groups) and many model versions  
 → more with higher resolution  
 → 21 MIPs, many experiments, larger ensembles  
 → Expectations:  
 Volume: 150 PB  
 Number of files: 280 Mio

## How to characterize the wide variety of models in CMIP6?

### → Routine Benchmarking and Evaluation Central Part of CMIP6 -

- **Evaluation tools** are provided such as the Earth System Model Evaluation Tool (ESMValTool, *Eyring et al., 2016*) the NCAR CVDP (*Phillips et al., 2014*) the PCMDI Metrics Package (PMP, *Gleckler et al., EOS, 2016*)
- will produce well-established analyses as soon as CMIP model output is submitted

## Broad Characterization of Model Behavior



Rel. space–time root-mean square error calculated from the 1980–2005 climatological seasonal cycle of the CMIP5 historical simulations. Blue/red shading indicating performance being better/worse than the median of all model results.

Pacific-Decadal Oscillation (PDO), 41 CMIP5 models and observations (upper left panel) for 1900–2005.

# Open questions: Big data & data sciences in climate sciences

## Hypothesis-driven Approach (with some data science)

Step 1: Analysis of changes – Evaluation of climate models

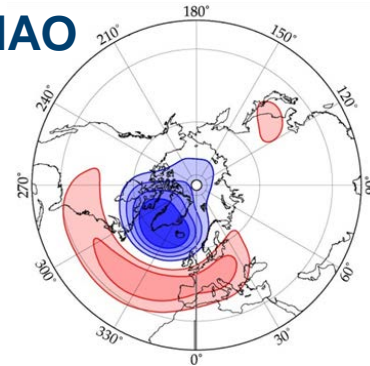
Step 2: Development of hypothesis

Step 3: Provision of evidence

→ New model experiments

→ New analysis

NAO



## Current Technological Approach

- Download of data from data centers
- Data analysis locally
- Software packages (MATLAB, R)
- Own software (FORTRAN, R) → use of libraries (NAG)



- Our current technological approach for climate data analysis will be probably not applicable for CMIP6 and other future modelling activities
  - How can we benefit from routine benchmarking and evaluation within CMIP6?
  - How can we perform data analysis remotely given special software needs?
  - How to reduce the analytical bottleneck in scientific data analysis?
  - How to visualize results (large ensembles)?
- **There is a need for theory-guided/hypothesis-driven data science methods that blend the power of big data analytics with the caution of scientific theory and first principles.** (Faghmous & Kumar, 2014)