



1 not uniform in time, there are two well known periods of acceleration: 1920-1940 and 1980-
2 2000, and two periods with stabilization of global mean temperature: 1950-1975 and 2000-2014.
3 The reason of this oscillatory behavior is still debated. In Wilcox et al (2013) it is shown that
4 period of climate stabilization in 1950-1975 can be connected with the increase of anthropogenic
5 SO₂ emission in Europe and North America as well as with stratospheric volcanic eruptions
6 (Bindoff et al 2013), while decrease of warming in 2000-2014 could be attributed to slowdown
7 of methane and tropospheric ozone concentration increase rate. On the other hand the ensemble
8 of CMIP5 model runs (with the all mentioned aspects of aerosol and greenhouse gas forcing
9 taken into account) continues to rise global temperature in 2000-2014 albeit with the slower rate
10 (Checa-Garcia et al (2016)).

11 Another point of view on this problem is that the acceleration and deceleration of global
12 warming could be a manifestation of internal climate variability with the time scale of 60-70
13 years (Meehl et al 2011). The Atlantic Multidecadal Oscillation (AMO) and Pacific Decadal
14 Oscillation (PDO) are the most known drivers of internal variability in the climate system on
15 multidecadal time scales. Indeed, Dong and McPhaden (2017) showed the importance of AMO-
16 like and PDO-like internal variability for local temperature in the North Atlantic and North
17 Pacific but questioned its ability to produce significant anomaly in global mean temperature. A
18 connected question is to what extent the observed long-term variability of the AMO and PDO
19 patterns is an internal process, or is forced by some external factors. There is some evidence
20 (Ting et al 2014) that negative values of the AMO index in 1950-1970 could be attributed to the
21 enhanced SO₂ emission in Europe and North America.

22 One of the most intriguing features of recent climate changes is a rapid decrease of Arctic
23 sea ice area since year 2000 coupled with strong Arctic warming. Similar Arctic warming was
24 also observed in the middle of the 20th century. The ensemble of CMIP5 models underestimates
25 the amount of sea ice loss in 2000s by a factor of two (Bindoff et al 2013, Stroeve et al 2012).
26 The INMCM4 climate model (Volodin et al 2013) participated in CMIP5 also strongly
27 underestimates Arctic sea ice extent loss in the beginning of the 21st century. On the other hand
28 the INMCM4 (and other CMIP5 models, Stroeve et al 2012) demonstrates loss of Arctic sea ice
29 of comparable magnitude at different times (in the middle of the 20th century for INMCM4).
30 Moreover, similar sea ice loss event was produced by INMCM4 during a preindustrial control
31 run. So, the question is to what extent the 21st century Arctic sea ice degradation is due to the
32 internal climate variability and whether the next generation of models with new CMIP6 forcing
33 recommendations is able to reproduce sea ice changes in the beginning of 21st century?

34 Regional climate changes during the last several decades also show some interesting
35 features. For example, in 2000-2014 there is almost no winter warming in the majority of Eurasia



1 with respect to the previous decades and even a small cooling was observed in some places. One
2 possible reason could be the response of atmospheric dynamics to Arctic sea ice loss (Overland
3 et al 2011). But this hypothesis is questioned by other studies (see McCusker et al (2016) as an
4 example).

5 The aim of this study is to analyze basic features of climate changes during the 1850-
6 2014. The data for the analysis (ensemble of seven historical runs) was produced by the new
7 climate model INM-CM5 being an incremental upgrade of the INMCM4. We are mainly
8 focusing on the question of how global mean surface temperature (GMST) changes are
9 reproduced with the new forcing protocols proposed for CMIP6 and how these changes are
10 connected with the reproduction of other features of the climate system mentioned above (i.e.
11 AMO and PDO variability etc).

12

13 2. Model and data.

14

15 The climate model INM-CM5 (Volodin et al 2017A; Volodin et al 2017B) was used in
16 this study. In the atmosphere, it has spatial resolution of 2×1.5 degrees in longitude and latitude,
17 and 73 levels in vertical, with the uppermost level at 0.2 hPa. In the oceanic block, spatial
18 resolution is 0.5×0.25 degrees and 40 levels in vertical. The model includes an interactive aerosol
19 block (Volodin and Kostrykin 2016), where concentrations of 10 aerosols are calculated. In
20 numerical experiments discussed below only the first aerosol indirect effect (the influence of
21 aerosol on cloud drop radius) is taken into consideration. Model description and analysis of
22 simulation of the present day climate can be found in Volodin et al (2017B).

23 Let us discuss now a climate change modeling experiment for years 1850-2014. Time
24 series of CO_2 , CH_4 , N_2O , O_3 , stratospheric volcanic sulfate aerosol concentration, solar constant
25 and solar spectrum, as well as anthropogenic emissions of SO_2 , black carbon and organic carbon
26 were prescribed as recommended for historical run of CMIP6 ([https://esgf-
27 node.llnl.gov/search/input4mips/](https://esgf-node.llnl.gov/search/input4mips/)). Seven model runs were started with different initial
28 conditions obtained from long preindustrial run, where all external forcings were prescribed at
29 the level of year 1850. The length of preindustrial run was several hundred years, so upper
30 oceanic layer was adjusted to atmospheric model conditions, but it is not the case for the deep
31 ocean. A small trend of model climate is visible because of deep ocean adjustment to upper
32 oceanic and atmospheric conditions – a common situation for simulation of historical climate
33 with present day climate models. The obvious reason for multiple integrations is to separate the
34 role of natural variability and external forcing in climate changes. When data of seven model
35 runs are consistent with each other, then one can expect that the phenomenon of interest is a



1 manifestation of (or response to) an external forcing. If there is a noticeable difference between
2 different model runs, then a role of natural variability is crucial. To estimate statistical
3 significance of near surface temperature trend, t-test at 99% level was used. Variance of 5 year
4 means was calculated from 1200 years of preindustrial run.

5 Observational data of GMST for 1850-2014 used for verification of model results were
6 produced by HadCRUT4 (Morice et al 2012). Monthly mean sea surface temperature (SST) data
7 ERSSTv4 (Huang et al 2015) are used for comparison of the AMO and PDO indices with that of
8 the model. Data of Arctic sea ice extent for 1979-2014 derived from satellite observations are
9 taken from Comiso and Nishio (2008). Stratospheric temperature trend and geographical
10 distribution of near surface air temperature trend for 1979-2014 are calculated from ERA Interim
11 reanalysis data (Dee et al 2011).

12

13 **3. Results.**

14

15 The most important measure of climate changes is the global mean surface temperature.
16 Observed GMST demonstrates the well known acceleration of warming in 1920-1940 and 1980-
17 2000 and small warming or even small cooling in 1945-1970 and 2000-2014. The ensemble of
18 CMIP5 models (Bindoff et al 2013) shows less significant slowdown in warming in 2000-2014.
19 In particular, the INMCM4 model (Volodin et al 2013) demonstrates gradual warming starting
20 from 1920.

21 With the new CMIP6 protocols all seven INM-CM5 model runs demonstrate fast
22 warming in 1980-2000 with the rate close to the observations and GMST stabilization in 2000-
23 2014 and 1950-1970 (Fig.1). The only significant difference in the new CMIP6 forcings at the
24 beginning of 21st century with respect to CMIP5 ones is the change in the Solar constant. Before
25 year 2000 CMIP5 and CMIP6 Solar constants show almost identical behavior (CMIP5 Solar
26 forcing can be found at <https://pcmdi.llnl.gov/mips/cmip5/forcing.html>). In 2001-2008 Solar
27 constant recommended for CMIP6 is about 0.3 W/m² lower than the one for CMIP5 (Fig.2). For
28 2009-2014 CMIP5 scenario suggested repetition of previous Solar cycle that gives the value of
29 the Solar constant almost 1 W/m² above the one recommended for CMIP6. An additional model
30 run with anthropogenic aerosol emissions fixed at the level of year 1850 shows gradual GMST
31 rise in 1950-1970 together with its stabilization 2000-2014 (not shown). The later fact supports
32 the hypothesis that correct reproduction of GMST changes in 2000-2014 is due to the corrected
33 CMIP6 treatment of the Solar constant.

34 Better representation of GMST stabilization in 1950-1970 (Fig.1) in simulations with
35 INM-CM5 with respect to the INMCM4 can be explained by incorporation of the new aerosol



1 block in the model that resulted in more sophisticated treatment of anthropogenic and volcanic
2 aerosol interaction with atmospheric radiation. Fast warming in 1920-1940 similar to the
3 observed one can be seen in four model runs, while other three runs show warming earlier or
4 later. These results suggest that the observed acceleration of warming in 1920-1940 is probably
5 due to combination of external forcing and natural variability.

6 Keeping in mind the arguments that the GMST slowdown in the beginning of 21st century
7 could be due to the internal variability of the climate system let us look at the behavior of the
8 AMO and PDO climate indices. Here we calculated the AMO index in the usual way, as the SST
9 anomaly in Atlantic at latitudinal band 0N-60N minus anomaly of the GMST. Model and
10 observed 5 year mean AMO index time series are presented in Fig.3. The well known oscillation
11 with a period of 60-70 years can be clearly seen in the observations. Among the model runs, only
12 one (dashed purple line) shows oscillation with a period of about 70 years, but without
13 significant maximum near year 2000. In other model runs there is no distinct oscillation with a
14 period of 60-70 years but period of 20-40 years prevails. As a result none of seven model
15 trajectories reproduces behavior of observed AMO index after year 1950 (including its warm
16 phase at the turn of the 20th and 21st centuries). One can conclude that anthropogenic forcing is
17 unable to produce any significant impact on the AMO dynamics as its index averaged over 7
18 realization stays around zero within one sigma interval (0.08). Consequently, the AMO dynamics
19 is controlled by internal variability of the climate system and cannot be predicted in historic
20 experiments. On the other hand the model can correctly predict GMST changes in 1980-2014
21 having wrong phase of the AMO (blue, yellow, orange lines on Fig.1 and 3).

22 More coherent behavior of model trajectories after year 1980 could be seen for the North
23 Atlantic (45-65N) temperature (Fig.4). Indeed the temperature deviates from its 1850-1899 mean
24 by 1.5 sigma in the early 2000s. The NA temperature index in the model shows notable
25 oscillations with periods of about 30-40 and 60-80 years (close to the 25 and 80 years for the
26 observations), three trajectories have correct strongly positive NA temperature anomalies in the
27 21st century.

28 Another important climate feature that could be responsible for the changes of the GMST
29 growth rate is the PDO measured by its index defined as normalized projection of the SST
30 anomaly on a specific pattern in the North Pacific at 20-60N. The 5 year average PDO index for
31 observations and model data is presented in Fig.5. For the observations, one can see maxima at
32 years 1930-1940 and 1980-1995 and a prolonged minimum during 1950-1975. None of the
33 model trajectories reflects observed time series of the PDO index by the same reasons discussed
34 earlier in the paragraph devoted to the AMO. Again model does not need correct PDO index
35 dynamics to predict GMST behavior.



1 One of the most intriguing observed features of ongoing climate changes is the fast
2 summer Arctic sea ice extent decrease in the beginning of 21st century. The ensemble of CMIP5
3 models underestimates the rate of decrease of Arctic summer ice area by the factor of two.
4 Model INMCM4 participated in CMIP5 also underestimates the value of Arctic sea ice extent
5 decrease significantly (Volodin et al 2013). In newly obtained INM-CM5 data (Fig.6) we see
6 qualitatively the same behavior of the Arctic sea ice as the average rate of the sea ice loss is
7 underestimated by the factor of the two to three. Though, in one model run (purple) the
8 magnitude of decrease is similar to the one in the observations (reduction from 7-7.5 million km²
9 in 1980s to 4-5.5 million km² in 2000s). In other runs Arctic sea ice loss is underestimated by
10 factor of 1.5-3, and in one run (green) one can even see some increase of Arctic sea ice area
11 during last decades. Our results suggest that the rapid decrease of Arctic sea ice extent near year
12 2000 was partially induced by an external forcing however the role of internal variability can be
13 very important (the range of the sea ice extent year-to-year variability could be estimated as 3.0
14 million km²).

15 The stratosphere is more sensitive to global changes than the troposphere. One can see in
16 the observations the stratospheric cooling by several degrees during the last decades. In the ERA
17 Interim reanalysis data (Fig.7) the global and annual mean temperature at 5 hPa in year 2014 is
18 3K lower than in year 1979. All model runs show gradual decrease of stratospheric temperature
19 during all period of historical run from 1850 to 2014, but the rate of decrease in 1979-2014 is
20 highest and equal 2.5K that is slightly below in absolute value than the observed one. This strong
21 decrease is consistent in all model runs, and is likely produced by combined effects of CO₂
22 increase and ozone decrease. Oscillations of global mean temperature at 5 hPa with period of 10-
23 12 years represent prescribed Solar cycle.

24 One of the characteristic features of climate changes in recent decades is a specific
25 geographical pattern of surface temperature trends. Figure 8 shows near surface air temperature
26 difference between 2000-2014 and 1985-1999 according to ERA Interim reanalysis and model
27 mean data. Statistical significance for model data was estimated using t-test, 99% confidence
28 level was used. Reanalysis data look noisier than model mean, but some observed features are
29 reproduced well by the model ensemble. Maximum warming up to 2.5 K appears in Arctic, in
30 Barentz and Kara seas; warming over high and midlatitudes in Eurasia and North America is
31 about 1 K, and lowest warming (or even cooling in reanalysis data) is located in Southern ocean.
32 Model warming is robust everywhere except some areas in Southern ocean, zone of deep
33 convection in North Atlantic and Zones of Gulf Stream and Kuroshio separation from the shore,
34 where natural variability is high. In Pacific, observed pattern connected with PDO, is not
35 reproduced in model mean data as well as in any individual model run. Figure 9 represents near



1 surface temperature model trend in two experiments (blue and green) having the maximum and
2 minimum Arctic warming. In the second one (green) there is no Arctic warming at all and even
3 some cooling, and warming over Eurasian and North American midlatitudes also is much
4 smaller than in model average data. Otherwise in the first case (blue) the Arctic warming in some
5 areas is as large as 7 K, and midlatitudinal warming over Eurasia and North America is higher
6 than in model average data.

7

8 **4. Conclusions**

9

10 Seven historical runs for 1850-2014 with the climate model INM-CM5 were analyzed. It
11 is shown that magnitude of the GMST rise in model runs agrees with the estimate based on the
12 observations. All model runs reproduce stabilization of GMST in 1950-1970, fast warming in
13 1980-2000 and a second GMST stabilization in 2000-2014 suggesting that the major factor for
14 predicting GMST evolution is the external forcing rather than system internal variability.
15 Numerical experiments with the previous model version (INMCM4) for CMIP5 showed
16 unrealistic gradual warming in 1950-2014. The difference between the two model results could
17 be explained by more accurate modeling of stratospheric volcanic and tropospheric
18 anthropogenic aerosol radiation effect (stabilization in 1950-1970) due to the new aerosol block
19 in INM-CM5 and more accurate prescription of Solar constant scenario (stabilization in 2000-
20 2014) in CMIP6 protocol. Four of seven INM-CM5 model runs simulate acceleration of
21 warming in 1920-1940 in a correct way, other three produce it earlier or later than in reality. This
22 indicates that for the year warming of 1920-1940 the climate system natural variability plays
23 significant role.

24 No model trajectory reproduces correct time behavior of AMO and PDO indices. Taking
25 into account our results on the GMST modeling one can conclude that anthropogenic forcing
26 does not produce any significant impact on the dynamics of AMO and PDO indices, at least for
27 the INM-CM5 model. In turns, correct prediction of the GMST changes in the 1980-2014 does
28 not require correct phases of the AMO and PDO as all model runs have correct values of the
29 GMST while in at least three model experiments the phases of the AMO and PDO are opposite
30 to the observed ones in that time. The North Atlantic SST time series produced by the model
31 correlates better with the observations in 1980-2014. Three out of seven trajectories have
32 strongly positive North Atlantic SST anomaly as the observations (in the other four cases we see
33 near-to-zero changes for this quantity).

34 The INM-CM5 has the same skill for prediction of the Arctic sea ice extent in 2000-2014
35 as CMIP5 models including INMCM4. It underestimates the rate of sea ice loss by a factor



1 between the two and three. In one extreme case the magnitude of this decrease is as large as in
2 the observations while in the other the sea ice extent does not change compared to the
3 preindustrial ages. In part this could be explained by the strong internal variability of the Arctic
4 sea ice but obviously the new version of INMCM model and new CMIP6 forcing protocol does
5 not improve prediction of the Arctic sea ice extent response to anthropogenic forcing.

6 Model reproduces several observed geographic features of near the surface air
7 temperature trend during the last decades, including Arctic amplification with maximum over
8 Barentz and Kara seas, warming of about 1K over Eurasian and North American midlatitudes
9 and weakest warming over Southern ocean. Case to case variability is very important here as
10 well.

11 The decrease of stratospheric temperature at 5 hPa during the period of 1979-2014 is
12 successfully reproduced by the model in all experiments. The magnitude of the temperature drop
13 is close to the one for ERA Interim data (2.5 and 3K).

14

15 **Acknowledgements**

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17 Academy of Sciences and supported by the Russian Science Foundation, grant 14-27-00126.
18 Climate model runs were produced at the supercomputer of the Joint Supercomputer Center of
19 the Russian Academy of Sciences and supercomputer Lomonosov at Moscow State University.

20

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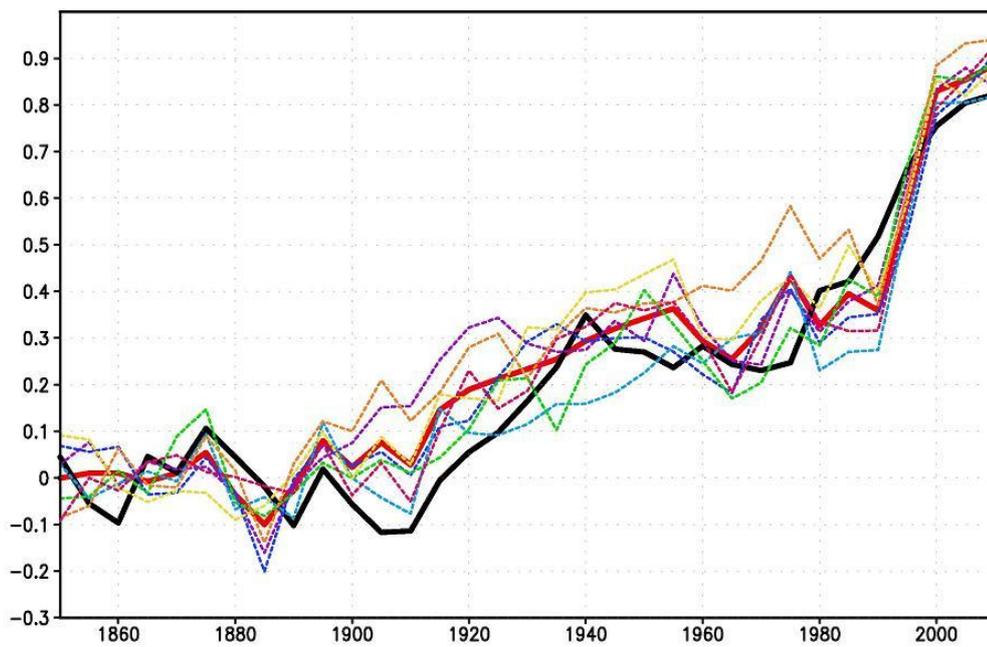
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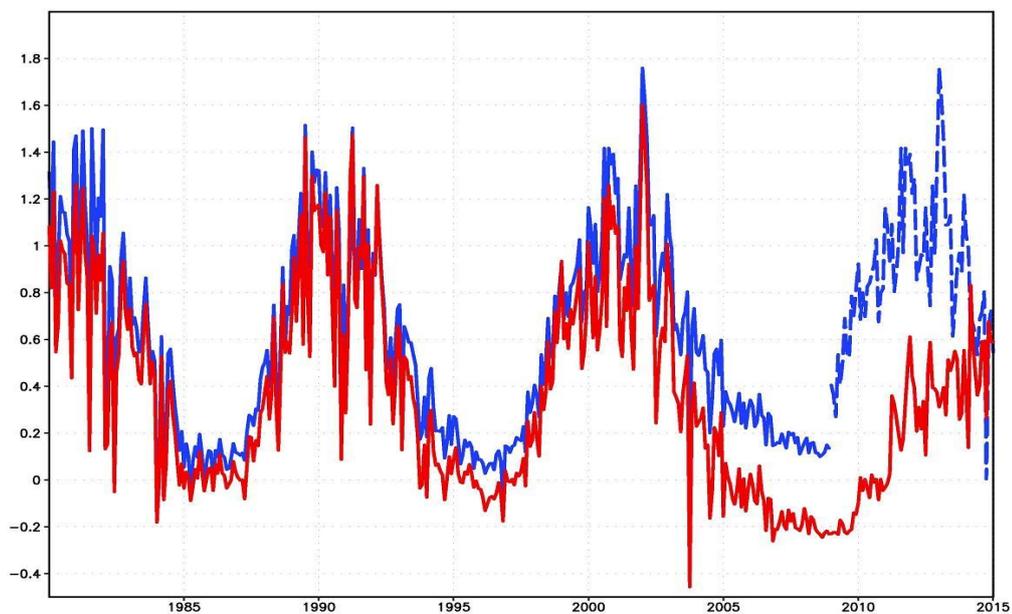


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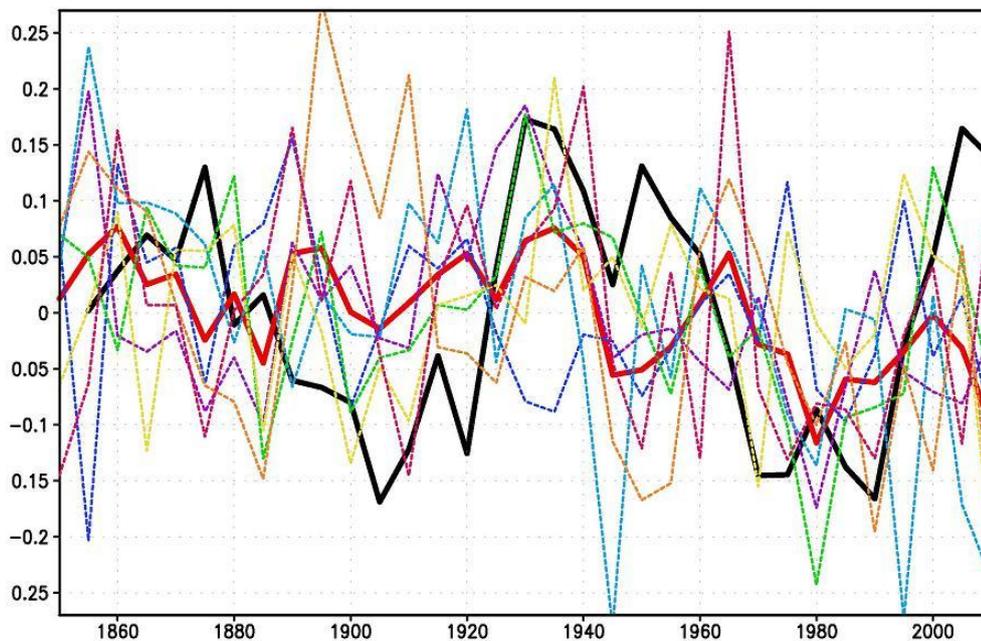
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Fig.1. Five year mean GMST (K) anomaly with respect to 1850-1899 for HadCRUTv4 (thick full black), model mean (thick full red). Dashed thin lines represent data of individual model runs: 1- purple, 2- dark blue, 3- blue, 4 – green, 5 – yellow, 6 – orange, 7 – magenta.



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Fig.2. Monthly mean Solar constant anomaly (W/m^2) with respect to 1882-1931 recommended for CMIP5 (blue; dashed line after year 2008 is the repetition of the data for 1998-2008) and for CMIP6 (red).



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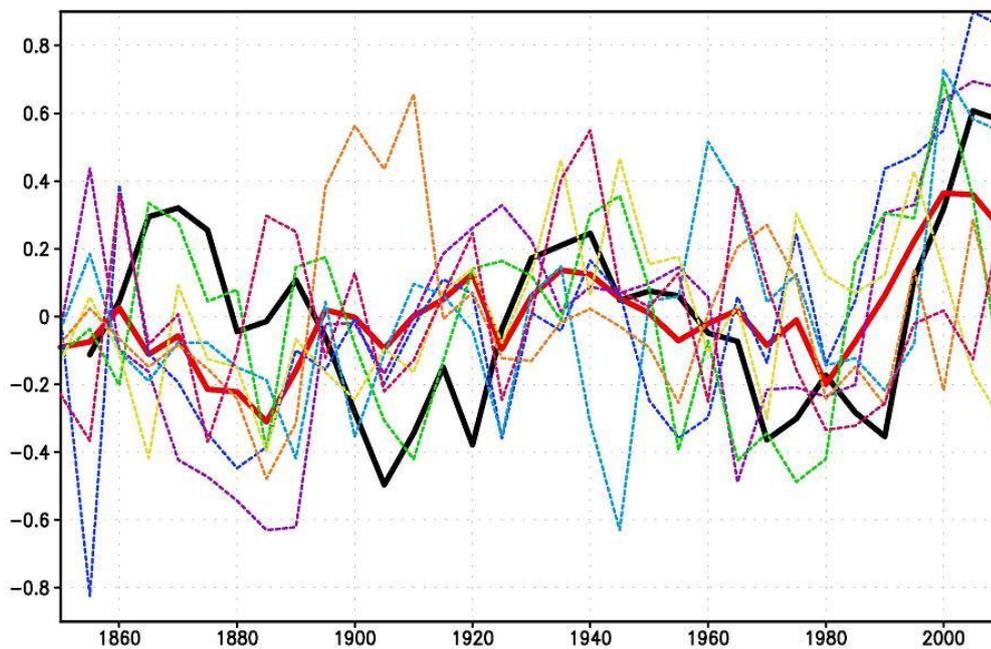
2 Fig.3. Five year mean AMO index (K) for ERSSTv4 data (thick full black), model mean (thick
3 full red). Dashed thin lines represent data of individual model runs. Colors correspond individual
4 runs as in Fig.1

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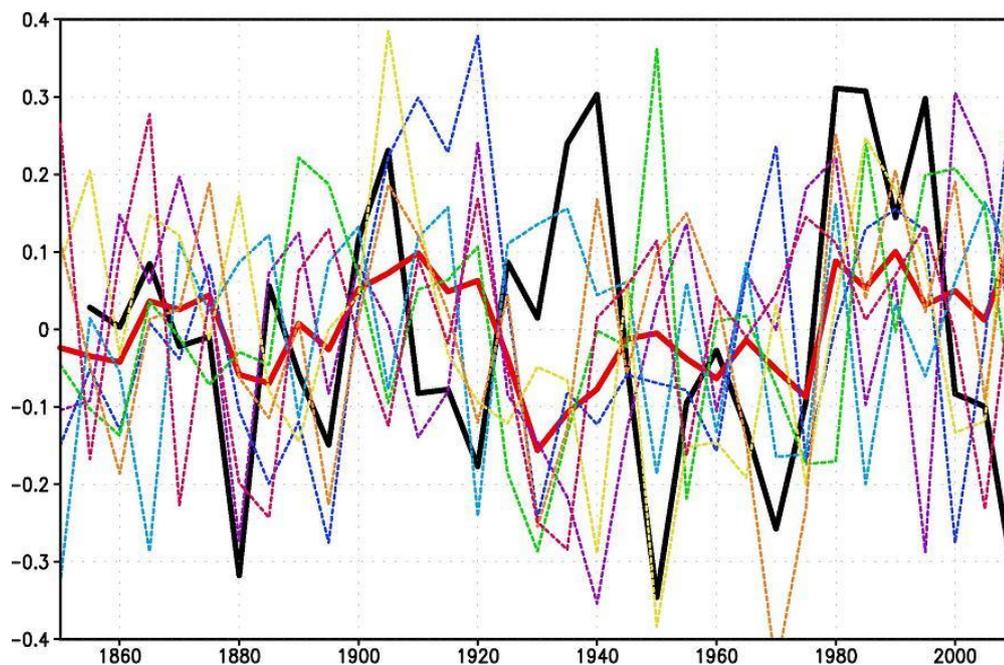
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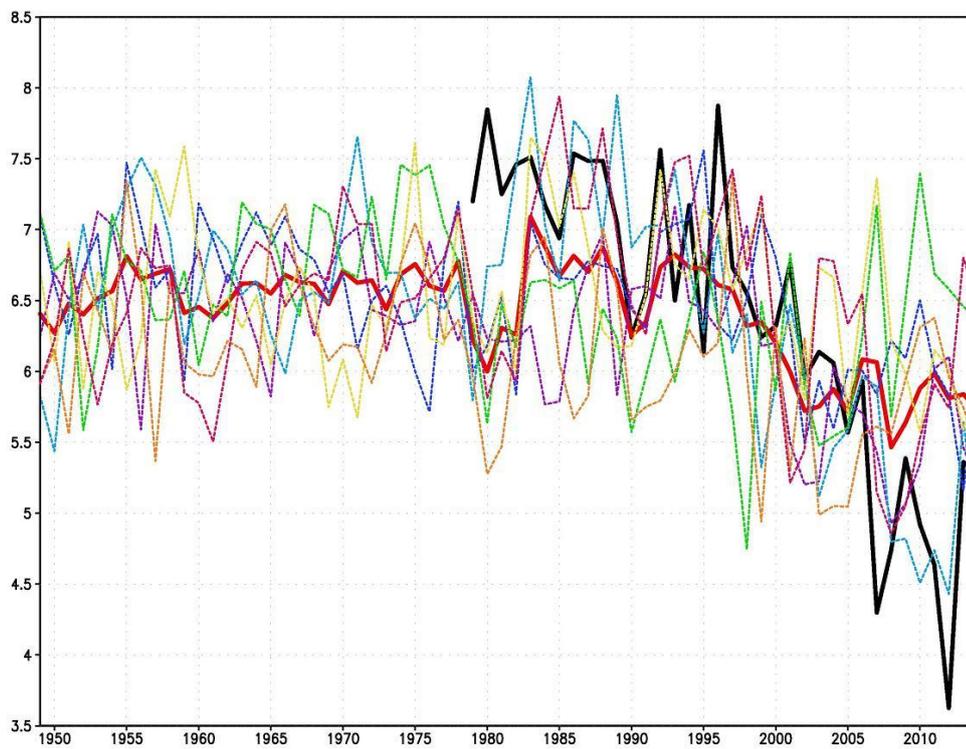
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2 Fig.4. Five year mean SST anomaly (K) with respect to 1850-1899 in North Atlantic (45N-65N)
3 for ERSSTv4 data (thick full black), model mean (thick full red). Dashed thin lines represent
4 data of individual model runs. Colors correspond individual runs as in Fig.1

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2 Fig.5. Five year mean PDO index (K) for ERSSTv4 data (thick full black), model mean (thick
3 full red). Dashed thin lines represent data of individual model runs. Colors correspond to
4 individual runs as in Fig.1.

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2 Fig.6. September Arctic sea ice extent (10^{12} m^2) for observations (Comiso and Nishio 2008)
3 (thick full black), model mean (thick full red). Dashed thin lines represent data of individual
4 model runs. Colors correspond individual runs as in Fig.1.

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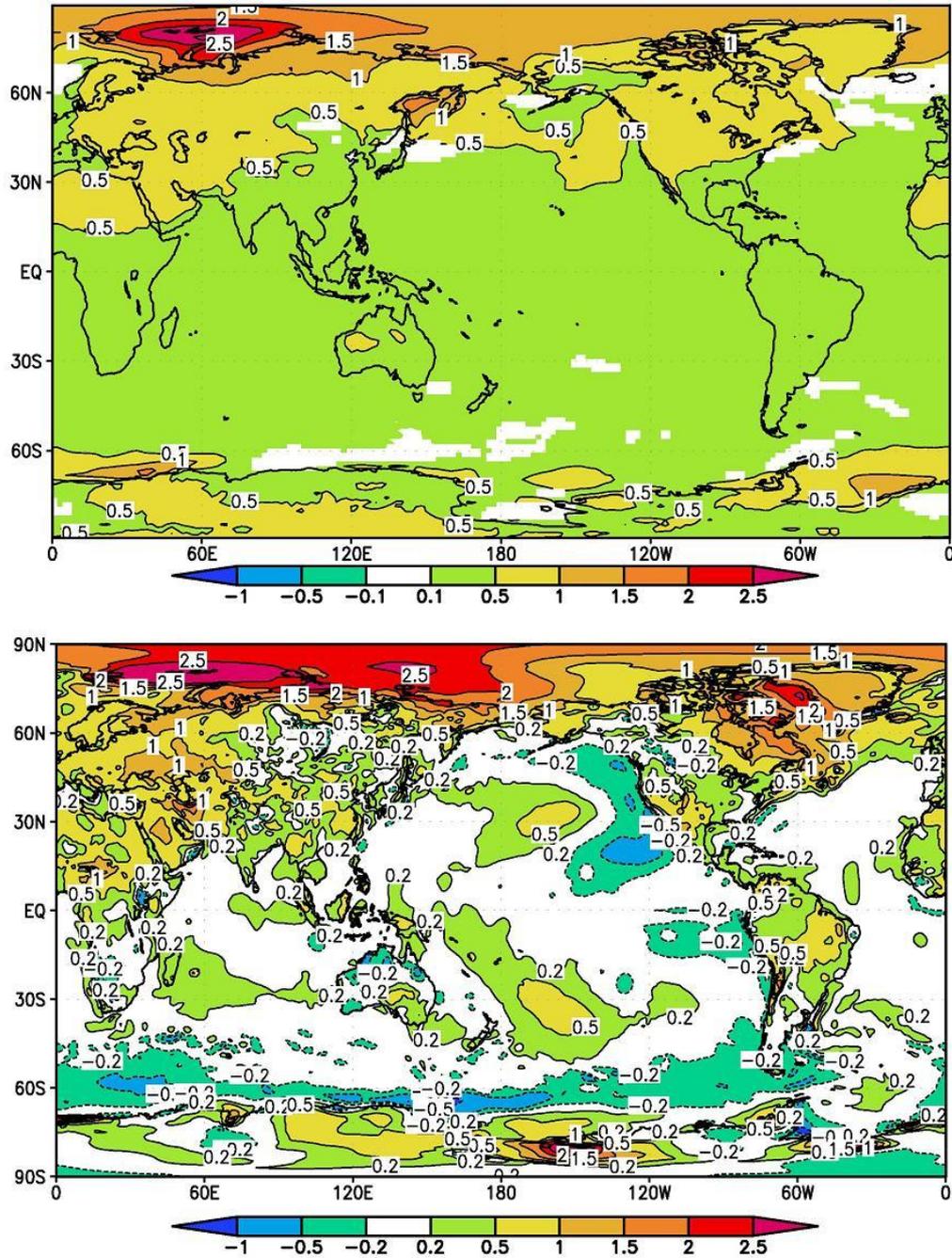


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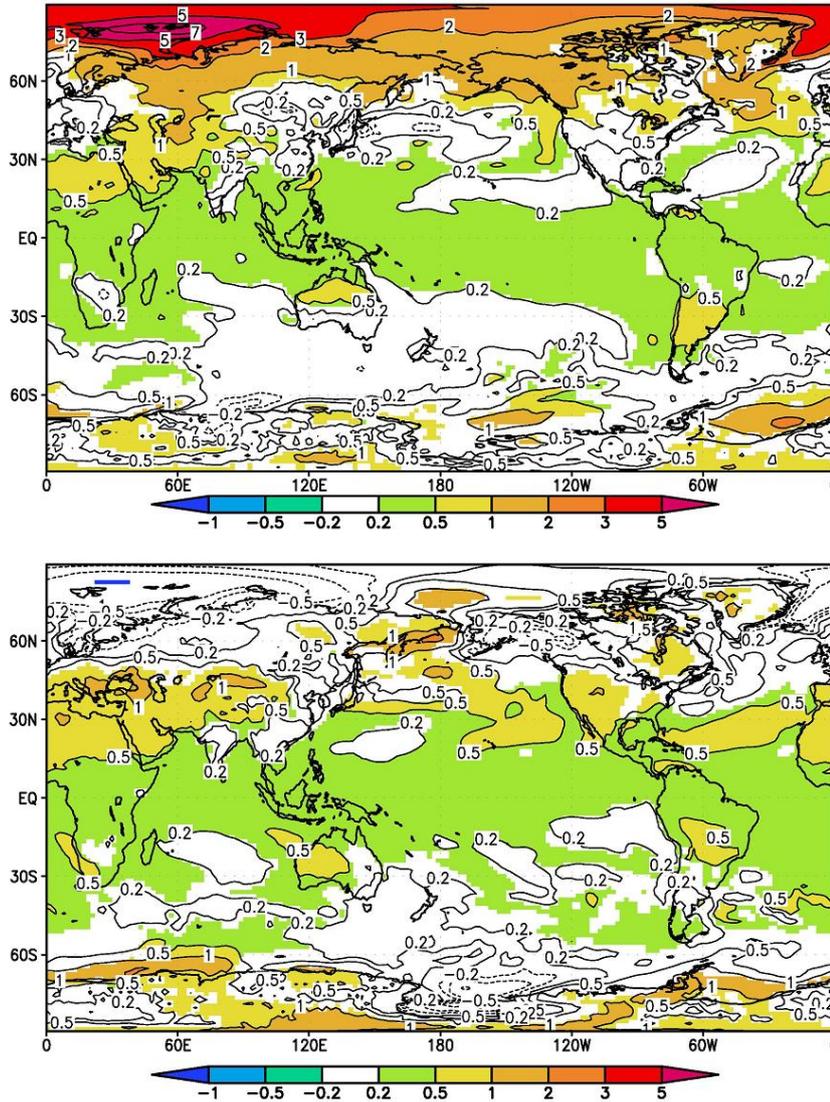
3 Fig.7. Annual mean global mean temperature (K) at 5 hPa for ERA Interim data (black) and
4 model data (dashed color lines).

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2 Fig.8. Annual mean near surface air temperature (K) in 2000-2014 minus 1985-1999 for model
3 mean data (top), shading represents 99% level of significance, and ERA Interim data (bottom).
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2 Fig.9. Annual mean near surface air temperature (K) in 2000-2014 minus 1985-1999 for model
3 run with highest (top) and lowest (bottom) warming in Arctic. Shading represents 99% level of
4 significance.