

1  
2  
3  
4  
5  
6  
7  
8  
9  
10  
11  
12  
13  
14

**Impacts of Atmospheric Temperature Changes  
on Tropical Cyclone Activity**

Submitted to *Journal of Climate*

27-July-2012

Gabriel A. Vecchi<sup>1</sup>, Stephan Fueglistaler<sup>2</sup>, Isaac M. Held<sup>1</sup>,  
Thomas R. Knutson<sup>1</sup>, Ming Zhao<sup>1,3</sup>

1. NOAA/GFDL, Princeton, NJ 08540 USA
2. Princeton University, Princeton, NJ 08540 USA
3. UCAR, Boulder, CO USA

**1 Abstract**

2 Aspects of the impacts of tropical temperatures in the upper troposphere (UT) and the  
3 tropical tropopause layer (TTL) on tropical cyclone (TC) activity are explored. UT and  
4 lower TTL cooling both lead to an overall increase in potential intensity (PI), while  
5 temperatures 70hPa and higher have negligible effect. Idealized experiments with a high-  
6 resolution global model show that cooler temperatures in the UT are associated with  
7 increases in global and North Atlantic TC frequency, but TC frequency changes in this  
8 model are not significantly affected by TTL temperature changes and do not scale with  
9 PI. Large uncertainties exist in the historical (1980-2008) UT and TTL trends, and the  
10 estimated impact of this uncertainty on TC activity is comparable to observed TC  
11 changes, impacting our confidence in assessments of the connection between hurricanes  
12 and climate.

13 Future projections of hurricane activity have been made with models that simulate the  
14 recent upward Atlantic TC trends while either assuming or simulating very different  
15 tropical temperature trends. This situation is illustrated, on the one hand, by several  
16 regional TC downscaling systems that successfully simulate these Atlantic trends while  
17 imposing the very strong UT and TTL trends of the NCEP Reanalysis, an outlier among  
18 observational estimates, and, on the other hand, by a high-resolution global model that  
19 simulate upward Atlantic TC frequency trends with a more nearly moist-adiabatic  
20 warming profile. Therefore, understanding the character of and mechanisms behind  
21 changes in UT and TTL temperature is crucial to understanding past and projecting future  
22 TC activity changes.

23 (248 words, max 260)

1

2 **1. Introduction:**

3 Understanding and modeling the links between climate and tropical cyclones (TCs) is  
4 a topic of substantial scientific interest (*e.g.*, Knutson *et al.* 2010), motivated in large part  
5 by the societal and economic impact of hurricanes (*e.g.*, Pielke Jr. *et al.* 2008;  
6 Mendelsohn *et al.* 2012; Peduzzi *et al.* 2012). Statistical, dynamical and hybrid statistical-  
7 dynamical models have all proved useful in the development of understanding of and  
8 predictive capability for hurricane activity changes (*e.g.*, Gray 1984; Elsner and Jagger  
9 2006; Oouchi *et al.* 2006; Camargo *et al.* 2007; Knutson *et al.* 2007, 2008; Emanuel *et al.*  
10 2008; Gualdi *et al.* 2008; LaRow *et al.* 2008; Swanson 2008; Zhao *et al.* 2009, 2010;  
11 Bender *et al.* 2010; Chen and Lin 2010; Smith *et al.* 2010; Vecchi *et al.* 2011, 2012;  
12 Villarini *et al.* 2011, 2012; Villarini and Vecchi 2012; Zhao and Held 2012). Most of  
13 these studies, and others (*e.g.*, Emanuel 2005; Vecchi and Soden 2007; Ramsay and  
14 Sobel 2011), have focused attention on the role of SST changes in directly or indirectly  
15 controlling past and future hurricane changes. However, processes controlling hurricane  
16 statistics may be impacted by changes in the atmosphere that are not closely tied to sea  
17 surface temperature (SST) (*e.g.*, Sugi and Yoshimura 2004; Yoshimura and Sugi 2005;  
18 Emanuel 2010; Held and Zhao 2011); in particular, upper atmospheric temperature  
19 changes could impact TC activity (*e.g.*, Emanuel 2010; Emanuel *et al.* 2012).

20 The dynamical and statistical-dynamical modeling studies that have explored the  
21 impact of climate variability and change on hurricanes to date can be categorized into two  
22 broad categories: 1) global models that are either forced by estimates of SST (*e.g.*, Oouchi  
23 *et al.* 2006; LaRow *et al.* 2008; Zhao *et al.* 2009, 2010), or coupled to ocean models (*e.g.*,

1 Gualdi *et al.* 2008; Smith *et al.* 2010); and 2) regional/limited-domain models that are  
2 forced by estimates of the large-scale 3D structure of the atmosphere, in addition to SST  
3 (*e.g.*, Knutson *et al.* 2007, 2008; Emanuel *et al.* 2008; Emanuel 2010). The changes to the  
4 3D structure of the atmosphere in the SST-forced and coupled global models emerges  
5 from the dynamical response of the model system to the imposed boundary conditions  
6 and forcings (which usually include changes in atmospheric composition, *e.g.* CO<sub>2</sub>  
7 concentrations, O<sub>3</sub>, volcanic aerosols); meanwhile, limited-domain models impose the  
8 evolution of the three-dimensional structure of the atmosphere. Thus, even though  
9 various global AGCMs may be forced with similar SSTs, their atmospheric temperature  
10 evolution need not be the same and may differ from those of observational estimates,  
11 including those used in limited domain models.

12 In particular, as was shown in Knutson *et al.* (2010), the observed history of Atlantic  
13 hurricane activity (including the multi-decadal trend) was recovered with comparable  
14 skill by the global SST-forced AGCM studies of Zhao *et al.* (2009,2010) and LaRow *et al.*  
15 *et al.* (2008), and the limited-domain studies of Knutson *et al.* (2007) and Emanuel *et al.*  
16 (2008). This similarity in historical North Atlantic hurricane hindcast skill emerges even  
17 though these limited-domain models were forced with the NCEP-NCAR Reanalysis  
18 (Kalnay *et al.* 1996), which, as generally recognized and as we explore further below  
19 (*e.g.*, Figure 1), has atmospheric temperature trends in the tropics and Atlantic that differ  
20 considerably from the tropical-mean moist-adiabatic warming and stratospheric cooling  
21 exhibited by the global models. Also, the temperature trends in the NCEP data are  
22 outliers when compared to other reanalyses and homogenized radiosonde data (*e.g.*,  
23 Pawson and Fiorino 1998). Therefore, to the extent that atmospheric temperature changes

1 have been important to hurricane activity, each limited-domain and global modeling  
2 system may have been achieving comparable hindcast skill for different reasons, and in  
3 response to large-scale changes that may have differed from those that occurred in the  
4 real climate system.

5 In this manuscript, specific attention is given to the possibility of an influence of  
6 temperatures in the tropical tropopause layer and lower stratosphere on hurricane  
7 intensity and trends (*e.g.*, Emanuel 2010; Emanuel *et al.* 2012). Temperatures in these  
8 layers can be influenced by trends in the strength of the stratospheric circulation and  
9 trends in radiatively active species such as ozone, both of which may be correlated with  
10 changes in the troposphere but cannot be understood in terms of changes of a moist  
11 adiabatic temperature profile in response to changes in SST (Fueglistaler *et al.*, 2009),  
12 which may not be captured by models that do not adequately resolve stratospheric  
13 dynamics and chemistry.

14 To what extent does the difference in the estimates of past tropical atmospheric  
15 temperature changes impact hurricane activity? And, at which pressure levels in the  
16 atmosphere are temperature/heating uncertainties/perturbations most influential? In the  
17 rest of the manuscript we attempt to address these questions. In Section 2 we describe our  
18 data sources and modeling framework. Section 3 focuses on the various estimates of the  
19 structure of atmospheric temperature change over the 1980-2008 period. Section 4  
20 assesses the role of different estimates of past atmospheric temperature changes on  
21 hurricane potential intensity. In Section 5 we explore the impact of idealized atmospheric  
22 heating profiles on hurricane activity in an AGCM. In Section 6 we summarize our  
23 results, offer some discussion of their implications and highlight key steps forward.

1

2 **2. Data and Methods:**3 *a. Observational analyses:*

4 We explore five different observationally-based estimates of the evolution of tropical  
5 atmospheric temperatures over the period 1980-2008. We use two products based  
6 entirely on homogenized radiosonde measurements of atmospheric temperature from  
7 select stations:

8 i) The United Kingdom's Meteorological Office's Hadley Centre Atmospheric  
9 Temperature (HadAT2) analysis (Thorne et al. 2005).

10 ii) Version 1.51 of the Radiosonde Innovation Composite Homogenization  
11 (RICH) made available at the University of Vienna (Haimberger *et al.* 2011)  
12 that is homogenized using information from neighboring radiosonde stations,  
13 and which provides an ensemble of 32 plausible records by varying  
14 parameters in the homogenization methodology.

15 The four other products used are observationally-constrained dynamical-model  
16 reanalyses:

17 i) The reanalysis produced by United States National Oceanic and Atmospheric  
18 Administration's National Center for Environmental Prediction (NCEP) and  
19 the United States National Center for Atmospheric Research (NCAR),  
20 referred to as the NCEP (Kalnay *et al.*, 1996). The data used are on a monthly,  
21 2.5°x2.5° grid, and archived at 17 standard pressure levels.

22 ii) The interim reanalysis produced by the European Centre for Medium Range  
23 Weather Forecasting (ECMWF), referred to as ERA-Interim (Dee *et al.*,

1           2011). The data used are on a monthly,  $1^\circ \times 1^\circ$  grid, and archived at 60 pressure  
2           levels between the surface and 0.1hPa.

3       iii)    The Modern-Era Retrospective Analysis for Research and Applications  
4           (MERRA; Rienecker *et al.* 2010), produced by the United States National  
5           Aeronautics and Space Administration (NASA). The data used are on a  
6           monthly,  $1.25^\circ \times 1.25^\circ$  longitude-latitude grid and 42 pressure levels between  
7           1000 and 0.1 hPa.

8       iv)    The NCEP Climate Forecast System Reanalysis (CFSR; Saha *et al.* 2010).  
9           The data used are on a monthly,  $2.5^\circ \times 2.5^\circ$  longitude-latitude grid and archived  
10          at standard pressure levels with additional intermediate pressure levels.

11       *b. Potential Intensity:*

12       Potential intensity (PI) is a theoretical upper bound on the intensity a TC can attain,  
13       given its environment. From observational estimates and models we compute PI as  
14       estimated by Bister and Emanuel (1998, 2002) based on monthly-mean values of sea  
15       surface temperature (SST), sea level pressure, and profiles of atmospheric temperature  
16       and humidity, using the Fortran code available here:

17       <ftp://texmex.mit.edu/pub/emanuel/TCMAX/>

18       *c. Model: HiRAM*

19       HiRAM is a global high resolution atmospheric model developed at GFDL with a  
20       goal of providing an improved simulation of the statistics of tropical storms. At about  
21       50km horizontal grid size, HiRAM forced by the observed sea surface temperatures  
22       (HadiSST) is found to be able to reproduce many aspects of the observed hurricane  
23       frequency variability for the past few decades, for which reliable observations are

1 available (Zhao *et al.* 2009). These include the geographical distribution of global  
2 hurricane tracks, the seasonal cycle, as well as the inter-annual variability and the decadal  
3 trend of hurricane frequency over multiple ocean basins. HiRAM has also been used to  
4 study hurricane seasonal forecasting in the N. Atlantic (Zhao *et al.* 2010; Vecchi *et al.*  
5 2011) and the results support a view that the overall activity of the Atlantic hurricane  
6 season has substantial predictability, if one can predict ocean temperatures. The historical  
7 simulations used for this study are also the integrations (3-member ensemble) for  
8 GFDL's participation in the CMIP5 high resolution time-slice simulations (Held *et al.*, in  
9 preparation) and the US CLIVAR Hurricane Working Group.

10

### 11 **3. Observed and Modeled Temperature Changes:**

12 There are some common features present in past upper troposphere (UT) and tropical  
13 tropopause layer (TTL) temperature changes in all the observational estimates and the  
14 HiRAM AGCM, though there are also important differences between different estimates.  
15 Figure 1 compares the time-evolution and 1980-2008 linear trends of atmospheric  
16 temperatures averaged over the peak season of northern hemisphere hurricane activity  
17 (July-October) from the observational estimates and the three-member ensemble mean of  
18 the HiRAM AMIP experiment, focussing on the tropical-average (30°S-30°N). Because  
19 hurricane changes in the tropical Atlantic have been a topic of particular interest in  
20 AGCM and limited domain hindcasts, we also explore values at the data point nearest the  
21 San Juan, Puerto Rico radiosonde station (66°W,18°N). Both in the tropical-mean and at  
22 San Juan, all the observational estimates show long-term cooling in the lower  
23 stratosphere (50hPa) and upper TTL (70hPa), punctuated by a notable warming in

1 response to the eruptions of Mts. El Chichón (1983) and Pinatubo (1991). The HiRAM  
2 model shows weaker 50-70hPa cooling than that of any of the observationally-based  
3 estimates, and it also shows reduced variability (HiRAM does not have a good  
4 representation of the quasi-biennial oscillation – QBO). In the lower TTL (100h-150hPa)  
5 HiRAM tends to track the non-NCEP observational relatively well, while NCEP is a  
6 considerable outlier.

7       The NCEP-NCAR Reanalysis stands out as an outlier among all the other  
8 observational estimates, by exhibiting very large cooling in the lower TTL and in the UT.  
9 In the lower TTL, the cooling in all products is smaller than in the lower  
10 stratosphere/upper TTL except NCEP, which shows a maximum cooling at 100hPa that is  
11 twice as large as the next largest cooling (CFSR). In the lower TTL the long-term cooling  
12 in HiRAM is smaller than that of most observational estimates (except ERA-Interim),  
13 and HiRAM shows no long-term change at 150hPa. The trends in HiRAM sit within the  
14 spread of the 32-member ensemble of trends from RICH between 850hPa and 150hPa  
15 (not shown). Meanwhile, the trends from NCEP are only within the RICH ensemble  
16 spread only at the 50hPa and 850hPa pressure levels (not shown).

17       In the UT there are also considerable differences across products. Most observational  
18 estimates have the sign of the long-term trend in temperature change from negative to  
19 positive between 150-200hPa, while that in NCEP has its zero crossing below 250hPa.  
20 The zero-crossing for HiRAM, at 150hPa, is higher than most observational estimates. In  
21 the upper and middle troposphere, the trends in HiRAM are within the range of the  
22 various observationally-constrained estimates. The NCEP and HadAT2 both show  
23 considerably less warming in the lower troposphere than do RICH, ERA-Interim, CFSR,

1 MERRA or HiRAM: in NCEP and HadAT2 the 1980-2008 tropospheric warming trends  
2 do not increase with height.

3 The results are similar over San Juan and in the tropical mean, with the outlier nature  
4 of NCEP in the lower TTL and UT more notable than in the tropical mean. Over the  
5 entire atmospheric column, the temperature trends in NCEP are for more cooling/less  
6 warming than those in HiRAM. In the upper TTL the trends from the NCEP are at the  
7 cold end of the spectrum of the various observational estimates, while those from the  
8 HiRAM AGCM are at the warm end.

9 In the lower TTL and UT the NCEP is an extreme outlier among the other products in  
10 its estimate of trends and in the temporal evolution of temperature. In the lower TTL, the  
11 trends in the HiRAM model are deviate from the non-NCEP datasets in the opposite  
12 direction to those of NCEP. In the UT, the temperature trends in HiRAM are less of an  
13 outlier, relative to the population – while the NCEP trends depart from the population as  
14 a whole down to ~300hPa.

15 Two divergent views of the past evolution of tropical atmospheric temperatures  
16 (NCEP and HiRAM) have been used in assessments of the changes of hurricane activity  
17 since 1980: the ZETAC regional model (Knutson *et al.* 2007, 2008; Garner *et al.* 2009)  
18 and the statistical-dynamical methodology of Emanuel *et al.* (2008) use the NCEP to  
19 drive their hurricane downscaling methodologies, while the HiRAM evolution emerges  
20 from the AGCM used to explore hurricane activity in Zhao *et al.* (2009, 2010), Zhao and  
21 Held (2011) and Held and Zhao (2011). Therefore, it is important to understand the  
22 extent to which these differences in the multi-decadal trends in atmospheric temperatures  
23 influence simulated hurricane activity.

1

2 **4. Atmospheric Temperature Changes and Potential Intensity:**

3 In this section we explore the impact of the differences in the temperature evolution  
4 between the NCEP and HiRAM on the Bister and Emanuel (2002) TC Potential Intensity  
5 (PI). Overall, the differences in atmospheric temperature trends between NCEP and  
6 HiRAM in the Atlantic main development region (MDR; 80°W-20°W, 10°N-25°N) are  
7 of a similar character than those averaged over the tropics. In addition to differences in  
8 temperature trends, NCEP and HiRAM have large differences in their trends in PI across  
9 the tropics (Figure 2.a-b). Trends in PI computed from NCEP are positive almost  
10 everywhere in the tropics, and exceed  $2.5 \text{ m}\cdot\text{s}^{-1}\cdot\text{decade}^{-1}$  over large areas of the tropics.  
11 Meanwhile, trends in PI computed from HiRAM are positive in as many places as they  
12 are negative, and the increases are almost everywhere less than  $2 \text{ m}\cdot\text{s}^{-1}\cdot\text{decade}^{-1}$ . There  
13 are also differences across the two datasets in the relationship between trends in PI and  
14 relative SST (the difference in local SST to the tropical average, which has been found to  
15 be a good predictor of the response of PI across a range of models – *e.g.*, Vecchi and  
16 Soden 2007; Gualdi *et al.* 2007; Xie *et al.* 2010; Ramsay and Sobel 2011; Camargo *et al.*  
17 2012). The patterns of PI trends exhibit some relationship to patterns of SST change in  
18 both datasets, but while in HiRAM the zero line of relative SST trends corresponds  
19 strongly to the zero line of PI trends, in the NCEP there are large areas with positive PI  
20 trends and negative relative SST trends (*e.g.*, the east Pacific).

21 The differences between the evolution of PI from the NCEP and HiRAM emerge  
22 clearly in their tropical-mean behavior (Figure 3.a), with the NCEP showing a clear  
23 increase but HiRAM no apparent change. The difference in PI trends emerges in the

1 regional, seasonal PI as well; for example, Fig. 3.b shows time series of PI from the  
 2 NCEP and HiRAM averaged over the Atlantic hurricane main development region  
 3 (MDR; 80°W-20°W, 10°N-25°N) over the hurricane season (June-November). Both  
 4 HiRAM and the NCEP exhibit increases in MDR PI over the 1980-2008 period – but the  
 5 increase in NCEP is more than twice as large as that in HiRAM.

6 What drives these differences in regional and global trends in PI between HiRAM  
 7 and the NCEP? PI depends on the surface enthalpy disequilibrium as well as the vertical  
 8 profile of temperature in the atmosphere, with larger enthalpy disequilibrium or cooler  
 9 UT/TTL temperatures leading to larger PI. Based on the differences in tropical  
 10 temperature trends in the NCEP and HiRAM, which show less warming/more cooling  
 11 aloft in the NCEP than in HiRAM, we hypothesize that a main source of the difference is  
 12 the difference in the tropical-mean temperature trends. To test this hypothesis we  
 13 recomputed PI from the NCEP data, after perturbing its temperature by replacing the  
 14 NCEP tropical-mean temperature linear least-squares trend at each point with that from  
 15 HiRAM. We scale the HiRAM temperature change by the SST trend in NCEP in order to  
 16 isolate the impact of differences in the profile of tropical-mean temperature trend on PI.  
 17 So, the modified temperature profile at each point is:

$$18 \quad T^*(x, y, p, t) = T_N(x, y, p, t) - \mathcal{L}_{\langle T_N \rangle}(p) \cdot t + \frac{\mathcal{L}_{\langle T_H \rangle}(p) \cdot \mathcal{L}_{\langle S_N \rangle}}{\mathcal{L}_{\langle S_H \rangle}} \cdot t \quad (\text{Eq. 1})$$

19 where:  $T^*$  is the modified temperature structure evolution,  $T_N$  and  $S_N$  are the NCEP  
 20 monthly atmospheric temperature and SST fields, respectively;  $T_H$  and  $S_H$  are the HiRAM  
 21 monthly atmospheric temperature and SST fields, respectively;  $\mathcal{L}_\xi$  is the slope of the  
 22 linear least-squares trend of quantity  $\xi$ ; and  $\langle \cdot \rangle$  is the tropical average of a quantity.  
 23 Differences between the NCEP PI evolution and that computed using the modified

1 temperature data reflect the impact of differences in the vertical structure of tropical-  
2 mean temperature trends between NCEP and HiRAM.

3 As can be seen in Figure 2.c, the PI trends over the globe are very similar between  
4 HiRAM and the modified NCEP, indicating that a large contribution to the regional  
5 differences in PI trend between the NCEP and HiRAM arise from differences in tropical-  
6 mean atmospheric temperature trends. There are still differences between the modified  
7 NCEP and HiRAM trends in certain regions: the modified NCEP has larger positive  
8 trends over the Indian Ocean; the modified NCEP has is largest Atlantic PI increase off  
9 the coast of South America, while in HiRAM it is off the coast of Africa. However, these  
10 regional differences are much smaller than the differences between the actual NCEP and  
11 HiRAM. It appears that the differences in regional PI trends between the modified NCEP  
12 and HiRAM are tied to differences in their relative SST trends; this HiRAM experiment  
13 is forced with the HadISSTv1 product (Rayner *et al.* 2003), while NCEP uses a different  
14 estimate of SST. For example, the relative warming of the Indian Ocean is more than  
15 twice as large in NCEP than in HadISSTv1 (used in HiRAM), and the relative warming  
16 of the Atlantic coincides with the location of the largest PI trends in each product. In fact,  
17 the zero line of the relative SST trend in NCEP corresponds very well with the zero line  
18 of PI trends in the modified NCEP, in contrast to that in the original NCEP (Fig. 2.a) but  
19 similar to HiRAM (Fig. 2.b).

20 The influence of the differences in tropical-mean temperature trends on PI can be  
21 seen through the blue lines in Figure 3, which show area averaged evolution and trends of  
22 PI computed from the modified NCEP temperature, which result in global and local  
23 trends that are much more similar between HiRAM and the modified NCEP – although

1 the year-to-year values can still differ. This analysis indicates that the dominant  
2 contribution to differences in trends in PI between the NCEP and HiRAM is the  
3 difference in the profile of tropical-mean atmospheric temperature trends in these two  
4 datasets.

5       So, from which atmospheric levels do the differences between HiRAM and NCEP  
6 temperature trends contribute the most to their differences in PI trends? In order to isolate  
7 these sources, we perform additional partial perturbations to the temperature profile in the  
8 NCEP, where we substitute  $T^*(x,y,p,t)$  at and above a certain pressure level only. The  
9 results from these partial perturbations are shown in Figure 4, focusing on the TTL (in  
10 blue), UT (dark orange) and rest of troposphere (light orange). More than half the  
11 difference between HiRAM and NCEP tropical-mean (and regional, not shown) PI trends  
12 is due to differences in temperature trends in the TTL, primarily from the large  
13 differences at 100hPa and 150hPa. Differences in temperature trends at and above 70hPa  
14 have a negligible impact on the differences in PI trends between NCEP and HiRAM. The  
15 differences in temperature trends in the troposphere also contribute a substantial amount  
16 to the differences in PI trends between the NCEP and HiRAM, with the NCEP cooling  
17 between 150hPa and 300hPa and HiRAM warming at those levels.

18       Therefore, to the extent that PI is an important control on hurricane intensity and,  
19 possibly, frequency as well, it is crucial to understand temperature trends that impact its  
20 evolution. We have found that differences between HiRAM and NCEP in their  
21 temperature trends at 70 hPa and higher up have little influence on trends in PI. However,  
22 temperature trends below that - roughly from 300 to 100hPa - have a substantial impact

1 on trends in PI, and correspondingly the uncertainties in tropical UT and TTL  
2 temperature trends are also a source of major uncertainty in trends of PI.

3

#### 4 **5. Atmospheric Heating Influences on HiRAM:**

5 Using the HiRAM AGCM we explore the impacts on TC activity of idealized heating  
6 anomalies that affect the mean tropical temperature profile and compare the model's  
7 response to those in PI computed from the model's temperature structure. We are  
8 encouraged to analyze HiRAM due to the quality of its simulation of TC genesis –  
9 climatology, variability, and trends (Zhao et al, 2009, 2010). In order to efficiently  
10 isolate the impact of atmospheric heating anomalies we build off of a control experiment  
11 forced with monthly climatological SST, with no interannual variability, computed from  
12 HadISST (Rayner *et al.* 2003) over 1981-2005. Four perturbation experiments were  
13 generated in which spatially uniform and time-invariant atmospheric cooling anomalies  
14 of different amplitudes were imposed over two different pressure ranges, one targeting  
15 the TTL and the other the UT, see Table 1.

16 The impact of the prescribed cooling anomalies on tropical-mean temperature,  
17 averaged over 20 years of model simulation, can be seen in the left panel of Figure 6. For  
18 reference, the right panel of Figure 5 shows the linear trends in tropical-mean temperature  
19 from NCEP, HiRAM-AMIP and their difference. All of the idealized heating experiments  
20 exhibit their largest temperature anomalies in the TTL, even when the heating was  
21 applied to the UT. The UT temperature anomalies in the experiments where the heating is  
22 applied to the TTL is considerably smaller than that in which the heating is applied to the  
23 UT. In this model, UT heating is has large impact on the TTL, but TTL heating does not

1 efficiently impact the UT. The atmospheric cooling perturbations also drive increases in  
2 precipitation and convective mass flux averaged through the tropics.

3 We expect that the cooling perturbations should lead to an increase in hurricane  
4 activity. This can be explored more explicitly and quantitatively by tracking hurricane-  
5 like vortices in the AGCM. Figure 6 shows the response of two metrics of global and  
6 Atlantic TC activity: i) fractional change in the number of hurricanes<sup>1</sup>, and ii) the change  
7 in the ratio of hurricanes to TCs (HU/TC ratio), which can be interpreted as a measure of  
8 storm intensity in this AGCM that is too coarse to capture the most intense TCs (*e.g.*,  
9 Zhao and Held 2010). In these HiRAM AGCM experiments, the amplitude of the cooling  
10 in the TTL is not a useful metric by which to discriminate the response of these TC  
11 metrics: the largest TTL cooling is in experiment B5, but it has the second weakest  
12 response in either hurricane measure, weaker than C1, which has the smallest cooling of  
13 the TTL. However, the mean cooling of the UT (150hPa-300hPa) is well correlated to the  
14 response of global and Atlantic frequency and HU/TC ratio (left panels Figure 6).

15 We also explore the relationship between PI changes and frequency changes in the  
16 right panels of Figure 6. PI changes in this set of experiments are well correlated with the  
17 change in HU/TC ratio (Fig. 6.f), showing a response of  $\sim 5\%$  per  $\text{ms}^{-1}$ . However, for  
18 hurricane frequency PI does not provide a clean description of the response across the  
19 four experiments – it can discriminate between the frequency response of the TTL and  
20 UT cooling experiments, but it cannot explain the differences in frequency response

---

<sup>1</sup> The maximum 10-meter wind speed obtained during the storm lifetime is used to define a TC or hurricane. Following the recommendation of Walsh (2007) for a model of this resolution, we reduce the standard criteria (17 m/s for TCs and 33m/s for hurricanes), by 10%. This adjustment has very little effect on the fractional changes in storm counts in the model.

1 across them. For example, for hurricane frequency experiments B4 and C1 have a similar  
2 PI change, but B4 does not have a significant change in frequency but C1 has a ~50%  
3 increase. The HiRAM AGCM shows sensitivity of TC intensity, but not of frequency, to  
4 PI changes.

5 The blue horizontal and green vertical lines place the model-estimated sensitivity of  
6 the various TC activity metrics in the context of the observed activity changes in the  
7 Atlantic and the differences between NCEP and HiRAM. Based on the HiRAM AGCM's  
8 sensitivity of frequency to UT temperature changes, the differences in UT temperature  
9 trends between NCEP and HiRAM project onto fractional changes that are comparable  
10 compared to the observed changes. That is, the differences between HiRAM and NCEP  
11 trends appear to be a first order effect relative to the observed TC frequency trends in the  
12 Atlantic in this AGCM. For the HU/TC ratio changes, the observed trends have been  
13 much smaller (indistinguishable from zero, but nominally -5% / 29 years – other  
14 observations measures of intensity show a clearer increase over this period; *e.g.*, Elsner  
15 *et al.* 2008) so the discrepancies between NCEP and HiRAM trends in their PI and UT  
16 temperature trends are proportionately large.

17

## 18 **6. Summary and Discussion:**

19 We have explored uncertainties in multi-decadal changes in atmospheric temperature,  
20 and their influence on TC PI and frequency. Over the period 1980-2008 multiple  
21 observational estimates agree that the troposphere has warmed and the stratosphere has  
22 cooled, but disagree on the magnitude of the tropospheric warming and its vertical  
23 structure, and on the character of temperature changes in them upper troposphere (UT)

1 and tropical tropopause layer (TTL). The large differences in temperature trends between  
2 different observational estimates, and their difference to the trend in the HiRAM AGCM,  
3 project onto uncertainties in TC metrics.

4 We have focused on the differences (and impacts) of the 1980-2008 trends in the  
5 NCEP-NCAR Reanalysis (NCEP) and of the SST-forced GFDL-HiRAM AGCM because  
6 the HiRAM model has been used to explore the sensitivity of hurricanes to climate (*e.g.*,  
7 Zhao *et al.* 2009, 2010) and the NCEP has been used as a forcing to limited-domain  
8 models used to understand the hurricane-climate connection (*e.g.*, Knutson *et al.* 2007,  
9 2008; Emanuel *et al.* 2008, 2012; Emanuel 2010). The trends from the NCEP-NCAR  
10 Reanalysis (NCEP) product deviate most strongly from the rest of the estimates explored,  
11 and show much stronger cooling of the TTL and upper troposphere than any other  
12 estimate, with cooling extending from the stratosphere to 300hPa. Overall, the tropical-  
13 mean and Atlantic temperature atmospheric temperature trends in HiRAM tend to be  
14 within the spread of the various non-NCEP estimates, while the trend in NCEP between  
15 300hPa and 100hPa is an extreme outlier among the various products explored (Figure 1).

16 Trends over 1980-2008 in Bister and Emanuel (1998, 2002) TC potential intensity  
17 (PI) also differ considerably between NCEP and GFDL-HiRAM (Section 4). The 1980-  
18 2008 trends in PI from GFDL-HiRAM closely follow relative SST (the difference  
19 between local and tropical-mean SST changes), in agreement with estimates of multi-year  
20 to centennial sensitivity of PI in coupled and atmospheric models (*e.g.*, Vecchi and Soden  
21 2007; Gualdi *et al.* 2008; Xie *et al.* 2010; Ramsay and Sobel 2011; Camargo *et al.* 2012).  
22 Meanwhile, because of a large tropical-mean increase in PI in NCEP, the PI trends in the

1 NCEP do not follow relative SST as closely, with large regions showing PI increases  
2 with relative SST decreases.

3         The differences in 1980-2008 trends in large-scale PI in NCEP and GFDL-  
4 HiRAM arise primarily because of differences in tropical-mean atmospheric temperature  
5 trends in both products, with the NCEP showing less warming/more cooling throughout  
6 the entire atmosphere. The principal contributors to the NCEP to GFDL-HiRAM  
7 difference in PI are temperature trends in the lower TTL (150 and 100hPa) and upper  
8 troposphere (between 300 and 150hPa), with differences in the lower troposphere leading  
9 to a relatively smaller (<20%) part of the PI differences. Temperature trend differences at  
10 and above 70hPa have a negligible impact on the difference of PI in these two products.  
11 The tropical UT and TTL cooling in NCEP leads to an overall increase in PI, while  
12 HiRAM shows no tropical-mean PI change over 1980-2008. The strong TTL and UT  
13 cooling in NCEP make trends in PI deviate from the tight relationship to relative SST that  
14 has been noted in other studies (*e.g.*, Vecchi and Soden 2007; Gualdi *et al.* 2008; Xie *et*  
15 *al.* 2010; Ramsay and Sobel 2011; Camargo *et al.* 2012). The tendency for NCEP to have  
16 a large tropical-mean trend in PI that has been noted in the literature (*e.g.*, Emanuel 2007,  
17 2010) arises primarily from the large cooling in the lower TTL and UT in NCEP (Figure  
18 3), features that are outliers among other observationally-based products (Figure 1).  
19 NCEP and HiRAM also use SST products that have exhibited different trends in tropical-  
20 mean and in patterns of SST (Figure 2; Vecchi and Soden 2007); so beyond differences  
21 associated with tropical-mean atmospheric temperatures, these differences in the patterns  
22 of SST change lead to differences in trends of regional PI between GFDL-HiRAM and  
23 NCEP.

1           In the idealized heating experiments with GFDL-HiRAM, a measure of intensity  
2 (the ratio of the number of hurricanes to TCs) shows a strong relationship to PI (Figure  
3 6.f). Intensity changes in this AGCM are more strongly connected to PI than are  
4 frequency changes. While this model is of insufficient resolution to represent the full  
5 spectrum of TC intensity, this ratio is reasonably well simulated in the model and so the  
6 model provides a crude estimate of intensity changes (Zhao and Held 2010).

7           Observational analyses have found a high correlation between SST changes in the  
8 tropical Atlantic and hurricane activity indices (*e.g.*, Elsner and Jagger 2006; Emanuel  
9 2005). However, observational correlations as high or higher have been found between  
10 hurricane activity and the weighted difference between Atlantic and tropical-mean SSTs  
11 (the SST changes in the Atlantic relative to the tropics, or “Relative SST”) by other  
12 studies (*e.g.*, Swanson 2007, 2008; Vecchi *et al.* 2008; Villarini *et al.* 2010, 2011.a, 2012;  
13 Villarini and Vecchi 2012). The physical basis for exploring relative SST as a predictor  
14 of hurricane activity is based on the tendency of free tropospheric temperature changes to  
15 follow those of tropical-mean SST (Sobel *et al.* 2002) or SSTs in the Indo-Pacific region  
16 where the bulk of tropical convection resides (Chiang and Sobel 2002; Tang and Neelin  
17 2004) as described by the Weak Temperature Gradient approximation (Sobel and  
18 Bretherton 2000). An Atlantic SST warming that is larger than that of the tropical  
19 average, with a tropospheric warming in the Atlantic that follows tropical-mean SST,  
20 would lead to a large-scale destabilization of the atmosphere in the Atlantic, to changes in  
21 the large-scale vorticity, shear and atmospheric humidity, as well as to increases in TC  
22 potential intensity (*e.g.*, Latif *et al.* 2007; Vecchi and Soden 2007; Gualdi *et al.* 2008;  
23 Sugi *et al.* 2009, 2012; Zhao *et al.* 2009; Xie *et al.* 2010; Zhao and Held 2011; Ramsay

1 and Sobel 2011; Camargo *et al.* 2012). Supporting the notion of relative SST as a  
2 predictor for Atlantic hurricane activity, dynamical modeling studies have found that the  
3 threshold for TC genesis under projected climate changes over the 21<sup>st</sup> century increases  
4 along with the overall tropical warming (*e.g.* Knutson *et al.* 2008), and the interannual,  
5 decadal and climate change response of North Atlantic TC frequency simulated with a  
6 across a range of dynamical frameworks is well explained by relative SST (*e.g.*, Vecchi  
7 *et al.* 2008; Sugi *et al.* 2009, 2012; Zhao *et al.* 2009, 2010; Vecchi *et al.* 2011; Villarini *et*  
8 *al.* 2011.a; Knutson *et al.* 2012; Zhao and Held 2012). Changes of tropical atmospheric  
9 temperatures that are relatively horizontally uniform and larger in the upper troposphere  
10 than at the surface (approximating a “moist adiabatic” warming profile) are crucial to the  
11 “relative SST” interpretation of hurricane activity changes. Therefore, the strong  
12 departure from approximately moist adiabatic warming seen in NCEP leads to a different  
13 interpretation of the role of tropical-mean warming on hurricane activity, in which  
14 tropical-mean changes are less effective at balancing the impact of local SST changes.

15       The multi-decadal evolution of atmospheric temperature and TC PI in the  
16 hindcast experiments with HiRAM (Zhao *et al.* 2009, 2010) differs considerably from  
17 that of NCEP, which was used in ZETAC (Knutson *et al.* 2007, 2008) and the  
18 downscaling studies of Emanuel *et al.* (2008; E08) – yet these three studies reported  
19 comparable hindcast skill in tropical Atlantic hurricane activity. Historical hindcasts with  
20 HiRAM (Zhao *et al.* 2009) and ZETAC (Knutson *et al.* 2007) over 1980-2006 reported  
21 linear trends in hurricane counts that were comparable to those observed (the ZETAC  
22 trends were slightly larger than those observed), as did the statistical-dynamical  
23 methodology of E08. This suggests that the sensitivity of Atlantic hurricane activity to

1 changes in atmospheric temperature in these three systems may differ. However, the  
2 response of future projections by these three systems shows comparable sensitivity to  
3 relative SST (Villarini *et al.* 2010; Knutson *et al.* 2012) – suggesting some level of  
4 commonality in their sensitivity to climate. Can these results be reconciled?

5       Analysis of an extended time series (1980-2008) of hurricane frequency from  
6 HiRAM and ZETAC suggests a path towards reconciling the sensitivities of those two  
7 systems. The year-to-year correlations of Atlantic hurricane frequency in HiRAM and  
8 ZETAC to observations are comparable to each other and comparable to the results  
9 described in the original papers (Zhao *et al.* 2009; Knutson *et al.* 2007) over a shorter  
10 record. In addition, the ensemble of linear trends in Atlantic hurricane from HiRAM  
11 compares well with observed over the longer 1980-2008 period, as did the shorter record  
12 in Zhao *et al.* (2009). However, the addition of 2007 and 2008 to the original ZETAC  
13 1980-2006 time series leads to an Atlantic hurricane trend that goes from being  
14 somewhat larger than that observed to more than twice that observed (we note that the  
15 2007 and 2008 integrations with ZETAC were not available at the time that Knutson *et*  
16 *al.* (2007, 2008) were written). The difference in the HiRAM and ZETAC trends over  
17 1980-2008 are qualitatively consistent with the expectation from the differences in the  
18 atmospheric temperature trends present in both systems, assuming they have similar  
19 sensitivities to atmospheric temperature change. Therefore, Atlantic hurricanes in  
20 HiRAM and ZETAC need not have fundamentally different sensitivities to climate.

21       Reconciling the behavior of HiRAM and E08 is more problematic. The E08  
22 methodology shows a strong sensitivity of hurricane frequency to temperature changes in  
23 the TTL (Emanuel *et al.* 2012); meanwhile, HiRAM shows strong sensitivity of global

1 and Atlantic hurricane frequency to temperature changes in the UT, but not to TTL  
2 changes (Section 5). We speculate that this difference in sensitivity to TTL temperatures  
3 is related to a distinct sensitivity of frequency to PI: in the E08 methodology PI is a  
4 primary thermodynamic constraint on storm genesis; while in the GFDL-HiRAM model  
5 there is no consistent emergent relationship between PI and genesis. Since PI is impacted  
6 by TTL temperature changes (Figure 4), frequency in the E08 methodology is sensitive to  
7 TTL changes. The relationship that sometimes appears between PI and frequency in  
8 HiRAM reflects a connection of both PI and genesis to upper tropospheric temperature  
9 (Figure 6.d-e) and to patterns of SST (Zhao *et al.* 2009, 2010; Vecchi *et al.* 2011, Held  
10 and Zhao 2011), rather than a direct connection of PI to genesis. In global atmospheric  
11 (*e.g.* Sugi *et al.* 2002, 2012; Held and Zhao 2011; Zhao and Held 2012) and coupled  
12 (*e.g.*, Gualdi *et al.* 2008) models hurricane frequency tends to scale with changes in large-  
13 scale ascent, for which tropospheric stability is of greater relevance than TTL  
14 temperatures. At this stage, it appears that HiRAM and the E08 downscaling technique  
15 have distinct sensitivity to climate; in fact, a downscale of the HiRAM model shown here  
16 using E08 technique does not recover the trends in Atlantic hurricane frequency that  
17 emerge in the HiRAM model (Emanuel *et al.* 2012). We speculate that the similar  
18 sensitivity to relative SST in future projections from HiRAM and E08 (*e.g.*, Villarini *et*  
19 *al.* 2010; Knutson *et al.* 2012) arises because GCM projections of the future have an  
20 approximately moist adiabatic warming of the troposphere, in which both large-scale  
21 stability and potential intensity tend to follow relative SST – so the differences in genesis  
22 sensitivity in HiRAM and E08 are masked. The differences in sensitivity of HiRAM and

1 E08 are only apparent when atmospheric temperature changes depart strongly from a  
2 moist adiabatic profile.

3 All of the observationally-based temperature products explored here, as well as  
4 HiRAM, show a cooling of the TTL across the tropics – though they disagree in its  
5 magnitude. If, as the E08 methodology indicates, TTL cooling acts to increase hurricane  
6 frequency, one would perhaps expect to have seen an increase in global mean frequency  
7 since 1980. However, the observed record shows a non-significant decrease in global  
8 hurricane frequency (Zhao *et al.* 2009; Maue 2011). Meanwhile, HiRAM (which does not  
9 show a sensitivity of frequency to TTL temperature) is able to recover the observed  
10 global-mean hurricane trends, along with the increases seen in the Atlantic and decreases  
11 seen in the East and West Pacific (Zhao *et al.* 2009). The observed history of global  
12 hurricanes may indicate that TTL temperature changes are not a robust influence on  
13 hurricane frequency.

14 A question arises from this comparison of HiRAM, ZETAC and E08: how can we  
15 discriminate between diverging sensitivities in models that perform comparably in  
16 hindcast mode? The good quantitative agreement between observed and modeled  
17 hurricane history was key to building confidence in each system's representation of the  
18 sensitivity of hurricanes to climate. However, given the different climate changes that  
19 drove hurricane changes that resemble observations in each system, it is no longer  
20 sufficient to explore hurricane hindcast skill in the Atlantic alone. A clear assessment of  
21 the plausibility of the different temperature trend profiles that either emerge from or are  
22 fed to these systems now becomes crucial.

1           Currently there is considerable uncertainty as to the character of past tropical  
2 atmospheric temperature trends, which does not allow the rejection of the hypotheses that  
3 tropical-mean UT temperatures have warmed at a reduced, equal or larger rate than  
4 tropical-mean SST since the late-1970s from direct temperature observations (Thorne *et al.*  
5 *al.* 2005, 2011; Sherwood *et al.* 2008; Santer *et al.* 2008). However, indirect evidence  
6 from observed changes to the structure of zonal-mean wind (Allen and Sherwood 2008)  
7 and the SST threshold for strong convection (Johnson and Xie 2010) suggests that the  
8 tropical troposphere may have warmed approximately moist-adiabatically. Since the  
9 trends in atmospheric temperature at and below 100hPa in HiRAM are within the spread  
10 of the non-NCEP reanalyses and the radiosonde-only products (and are largely within the  
11 32-member spread of RICH). On the other hand, the large discrepancies in trends of  
12 lower TTL and UT temperature between NCEP and the rest of the estimates explored  
13 here (Figure 1) suggest to us that it is unlikely to be an accurate representation of the  
14 multi-decadal trends in tropical atmospheric temperature that appear so influential to  
15 hurricane activity.

16           These results highlight the need to understand the mechanisms behind the history  
17 of tropical atmospheric temperature changes. In particular, if the tropical-mean  
18 troposphere has not been warming like GCMs between 300 hPa and 100 hPa, there are  
19 implications to the evolution of TC potential and actual intensity. According to the GCM  
20 utilized here, if there are departures from the moist adiabat below 150hPa, this could  
21 influence TC frequency. Currently, most dynamical model projections of the 21<sup>st</sup> century  
22 show something resembling moist adiabatic warming in the tropics, but if the models are  
23 deficient in a process, or a key forcing has been neglected, the likely future evolution of

1 TC statistics may differ from these projections. If current models overestimate the  
2 warming of the upper troposphere and lower TTL in response to increasing CO<sub>2</sub>, we  
3 would expect that current projections of 21<sup>st</sup> century hurricane activity underestimate the  
4 potential for increases in frequency and intensity of TCs. On the other hand, if the  
5 underestimate of TTL cooling of the was due to misrepresentation of the impact of  
6 stratospheric ozone decreases, which are expected to recover over the coming century  
7 (WMO 2011), we would expect that projections for the coming century may overestimate  
8 the potential increases of TC intensity by not allowing for a “rebound” from ozone  
9 recovery.

10

**11 Acknowledgments:**

12 We are grateful to Mike Winton and Massimo Bollasina for comments and suggestions,  
13 and Peter Thorne and John Lanzante for discussions on atmospheric temperature records.

14 We thank ECMWF for providing the ERA-Interim data. MERRA data used in this  
15 study/project have been have been provided by the Global Modeling and Assimilation  
16 Office (GMAO) at NASA Goddard Space Flight Center through the NASA GES DISC  
17 online archive. We thank NCEP and NCAR for providing the CFSR data, available at  
18 <http://dss.ucar.edu/pub/cfsr.html>. HADAT2 temperature data was downloaded from  
19 <http://www.metoffice.gov.uk/hadobs/hadat/>. RICH temperature data was downloaded  
20 from <http://www.univie.ac.at/theoret-met/research/raobcore/>.

21

22

23

1 **References:**

2 Allen, R.J., and S.C. Sherwood, 2008: Warming maximum in the tropical upper  
3 troposphere deduced from thermal winds. *Nature Geosci.*, **1**, 399-403.

4 Bender, M.A., T.R.Knutson, R.E.Tuleya, J.J. Sirutis, G.A.Vecchi, S.T. Garner, and I.M.  
5 Held, 2010: Model impact of anthropogenic warming on the frequency of intense  
6 Atlantic hurricanes. *Science* **327**, 454–458.

7 Bister, M., and K. A. Emanuel 1998: Dissipative heating and hurricane intensity. *Meteor.*  
8 *Atmos. Phys.*, **65**, 233–240.

9 ———, and ———, 2002: Low frequency variability of tropical cyclone potential intensity.  
10 1. Interannual to interdecadal variability. *J. Geophys. Res.*, **107**, 4801,  
11 doi:10.1029/2001JD000776.

12 Camargo, S. J., K. A. Emanuel, and A. H. Sobel, 2007: Use of a genesis potential index  
13 to diagnose ENSO effects on tropical cyclone genesis. *J. Climate*, **20**, 4819–4834.

14 ———, M. Ting, and Y. Kushnir, 2012: Influence of local and remote SST on North  
15 Atlantic tropical cyclone potential intensity. *J. Climate* (submitted).

16 Chen, J. H., and S. J. Lin, 2011: The remarkable predictability of inter-annual variability  
17 of Atlantic hurricanes during the past decade. *Geophysical Research Letters*, 38  
18 (L11804), doi:10.1029/2011GL047629.

19 Chiang, J.C. H., A.H. Sobel, 2002: Tropical Tropospheric Temperature Variations  
20 Caused by ENSO and Their Influence on the Remote Tropical Climate. *J.*  
21 *Climate*, **15**, 2616–2631.

- 1 Dee, D.P. and coauthors, 2011: The ERA-Interim reanalysis: configuration and  
2 performance of the data assimilation system, *Q. J. R. Meteorol. Soc.*, **137**, 553-  
3 597.
- 4 Emanuel, K., 2005: Increasing destructiveness of tropical cyclones over the past 30 years.  
5 *Nature*, **436**, 686–688.
- 6 ———, 2010: Tropical cyclone activity downscaled from NOAA-CIRES Reanalysis, 1908-  
7 1958. *Journal of Advances in Modeling Earth Systems* **2**, 1-12, 2010.
- 8 ———, R. Sundararajan, and J. Williams, 2008: Hurricanes and global warming—Results  
9 from downscaling IPCC AR4 sim- ulations. *Bull. Amer. Meteor. Soc.*, **89**, 347–  
10 367.
- 11 ———, S. Solomon, D. Folini, S. Davis, and C. Cagnazzo, 2012: Influence of Tropical  
12 Tropopause Layer Cooling on Atlantic Hurricane Activity. *J. Climate* (submitted).
- 13 Fueglistaler, S., A.E. Dessler, T. Dunkerton, I. Folkins, Q. Fu, P.W. Mote, 2009: Tropical  
14 Tropopause Layer, *Rev. Geophys.*, **47**, RG1004, doi:10.1029/2008RG000267.
- 15 Gray, W. M., 1984: Atlantic seasonal hurricane frequency. Part I: El Niño and 30 mb  
16 quasi-biennial oscillation influences, *Monthly Weather Review*, **112**, 1649-1668.
- 17 Gualdi, S., E. Scoccimarro, and A. Navarra, 2008: Changes in tropical cyclone activity  
18 due to global warming: Results from a high-resolution coupled general circulation  
19 model. *J. Climate*, **21**, 5204–5228.
- 20 Haimberger, L., C. Tavalato, and S. Sperka, 2012: Homogenization of the global  
21 radiosonde temperature dataset through combined comparison with reanalysis  
22 background series and neighboring stations. *J. Climate* (in press).

- 1 Held, I.M., and M. Zhao, 2011: The response of tropical cyclone statistics to an increase  
2 in CO<sub>2</sub> with fixed sea surface temperatures. *J. Climate*, **24**, 5353–5364.
- 3 Jarvinen, B. R., C. J. Neumann, and M. A. S. Davis, 1984: A tropical cyclone data tape  
4 for the North Atlantic Basin, 1886–1983: Contents, limitations, and uses. Tech.  
5 Memo. NWS NHC 22, National Oceanic and Atmospheric Administration, 24 pp.
- 6 Johnson, N.C., and S.-P. Xie, 2010: Changes in the sea surface temperature threshold for  
7 tropical convection. *Nature Geosci.*, **3**, 842-845, doi:10.1038/ngeo1008
- 8 Kalnay and coauthors, 1996: The NCEP/NCAR 40-Year Reanalysis Project. *Bull. Amer.*  
9 *Meteorol. Soc.*, **77**(3), 437-471.
- 10 Knutson, T.R., J. J. Sirutis, S. T. Garner, I. Held, and R. E. Tuleya, 2007: Simulation of  
11 recent increase of Atlantic hurricane activity using an 18-km-grid regional model.  
12 *Bull. Amer. Meteor. Soc.*, **88**, 1549–1565.
- 13 ———, ———, ———, G. A. Vecchi, and I. Held, 2008: Simulated reduction in Atlantic  
14 hurricane frequency under twenty-first- century warming conditions. *Nat. Geosci.*,  
15 **1**(6), 359–364.
- 16 ———, *et al.*, 2010: Tropical cyclones and climate change. *Nature Geoscience* **3**, 157–163.
- 17 ———, *et al.*, 2012: Dynamical Downscaling Projections of Late 21st Century Atlantic  
18 Hurricane Activity: CMIP3 and CMIP5 Model-based Scenarios. *J. Climate*  
19 (submitted)
- 20 Landsea, C.W., G.A. Vecchi, L. Bengtsson, and T.R. Knutson, 2009: Impact of Duration  
21 Thresholds on Atlantic Tropical Cyclone Counts. *J. Climate* doi:  
22 10.1175/2009JCLI3034.1.

- 1 LaRow, T. E., Y. K. Lim, D. W. Shin, E. P. Chassignet, and S. Cocke, 2008: Atlantic  
2 basin seasonal hurricane simulations. *J. Climate*, **21**, 3191–3206.
- 3 Maue, R. N., 2011: Recent historically low global tropical cyclone activity. *Geophys.*  
4 *Res. Lett.*, **38**, L14803, doi:10.1029/2011GL047711
- 5 Mendelsohn, R., K. Emanuel, S. Chonabayashi, and L. Bakkensen, 2012: The impact of  
6 climate change on global tropical cyclone damage, *Nature Climate Change*, **2**, 205-  
7 209.
- 8 Oouchi, K., J. Yoshimura, H. Yoshimura, R. Mizuta, S. Kusumoki, and A. Noda, 2006:  
9 Tropical cyclone climatology in a global warming climate as simulated in a 20-km-  
10 mesh global atmospheric model: Frequency and wind intensity analysis. *Journal of*  
11 *the Meteorological Society of Japan* **84**, 259–276.
- 12 Pawson, S., and M. Fiorino, 1998: A comparison of reanalyses in the tropical  
13 stratosphere. Part I: thermal structure and the annual cycle, *Climate Dynamics*, **14**,  
14 631-644.
- 15 Peduzzi, P., B. Chatenoux, H. Dao, A. De Bono, C. Herold, J. Kossin, F. Mouton, and O.  
16 Nordbeck, 2012: Global trends in tropical cyclone risk, *Nature Climate Change*,  
17 **2**, 289-294.
- 18 Pielke, R. A. Jr and coauthors, 2008: Normalized hurricane damages in the United States:  
19 1900–2005 *Nat. Hazard. Rev.*, **9**, 29–42.
- 20 Ramsay, H. A., and A. H. Sobel, 2011: Effects of relative and absolute sea surface  
21 temperature on tropical cyclone potential intensity using a single-column model.  
22 *J. Climate*, **24**, 183–193.

- 1 Rayner, N. A., D. E. Parker, E. B. Horton, C. K. Folland, L. V. Alexander, D. P. Rowell,  
2 E. C. Kent, and A. Kaplan, 2003: Global analyses of sea surface temperature, sea  
3 ice, and night marine air temperature since the late nineteenth century. *J.*  
4 *Geophys. Res.*, **108**, 4407, doi:10.1029/2002JD002670.
- 5 Rienecker, M. and coauthors, 2011: MERRA: NASA's Modern-Era Retrospective  
6 Analysis for Research and Applications, *J. Clim.*, **24**, 3624-3648, doi:  
7 10.1175/JCLI-D-11-00015.1.
- 8 Saha, S., and coauthors, The NCEP Climate Forecast System Reanalysis, 2010: *Bull.*  
9 *Amer. Meteor. Soc.*, **91**(8), 1015-1057, doi:10.1175/2010BAMS3001.1.
- 10 Santer, B. D., and Coauthors, 2005: Amplification of surface temperature trends and  
11 variability in the tropical atmosphere. *Science*, **309**, 1551–1556.
- 12 ———, and Coauthors, 2008: Consistency of modelled and observed temperature trends in  
13 the tropical troposphere. *Int. J. Climatol.*, DOI: 10.1002/joc.1756.
- 14 Sherwood, S.C., C.L. Meyer, and R.J. Allen, 2008: Robust Tropospheric Warming  
15 Revealed by Iteratively Homogenized Radiosonde. *J. Climate*, **21**, 5336-5350.
- 16 Smith, D. M., R. Eade, N. J. Dunstone, D. Fereday, J. M. Murphy, H. Pohlmann, and A.  
17 A. Scaife,, 2010: Skillful multi-year predictions of Atlantic hurricane frequency,  
18 *Nature Geoscience*, **3**, 846-849.
- 19 Smith, T. M., R. W. Reynolds, T. C. Peterson, and J. Lawrimore, 2008: Improvement to  
20 NOAA's historical merged land–ocean surface temperature analysis (1880–2006).  
21 *J. Climate*, **21**, 2283–2296.
- 22 Sobel, A. H., and C. S. Bretherton, 2000: Modeling tropical precipitation in a single  
23 column. *J. Climate*, **13**, 4378–4392.

- 1 ———, I. M. Held, and C. S. Bretherton, 2002: The ENSO signal in tropical tropospheric  
2 temperature. *J. Climate*, **15**, 2702–2706.
- 3 Sugi, M., and J. Yoshimura, 2004: A mechanism of tropical precipitation change due to  
4 CO<sub>2</sub> increase. *J. Climate*, **17**, 238–243.
- 5 ———, H. Murakami, and J. Yoshimura, 2009: A reduction in global tropical cyclone  
6 frequency due to global warming. *SOLA*, **5**, 164–167.
- 7 ———, ———, and ———, 2012: On the Mechanism of Tropical Cyclone Frequency Changes  
8 due to Global Warming. *J. Meteorol. Soc. Japan*, **90A**, 397–408.
- 9 Swanson, K. L., 2007: Impact of scaling behavior on tropical cyclone intensities,  
10 *Geophys. Res. Lett.*, **34**, L18815, doi:10.1029/2007GL030851.
- 11 ———, 2008: Nonlocality of Atlantic tropical cyclone intensities. *Geochemistry  
12 Geophysics Geosystems* **9**, Q04V01, doi:10.1029/2007GC00184.
- 13 Thorne, P. W., D. E. Parker, S.F.B. Tett, P.D. Jones, M. McCarthy, H. Coleman, and P.  
14 Brohan, 2005: Revisiting radiosonde upper-air temperatures from 1958 to 2002,  
15 *J. Geophys. Res.*, **110**, D18105, doi:10.1029/2004JD005753, 2005.
- 16 ———, P. Brohan, H.A. Titchner, M.P. McCarthy, S.C. Sherwood, T.C. Peterson, L.  
17 Haimberger, D.E. Parker, S.F.B. Tett, B.D. Santer, D.R. Fereday, and J.J.  
18 Kennedy, 2011: A quantification of uncertainties in historical tropical  
19 tropospheric temperature trends from radiosondes, *J. Geophys. Res.*, **116**,  
20 D12116, doi:10.1029/2010JD015487.
- 21 Vecchi, G. A., and B. J. Soden, 2007: Effect of remote sea surface temperature change on  
22 tropical cyclone potential intensity. *Nature*, **450**, 1066–1071.

- 1 —, K.L. Swanson, and B.J. Soden 2008: Whither Hurricane Activity? *Science* **322**  
2 (5902), 687-689.
- 3 —, M. Zhao, H. Wang, G. Villarini, A. Rosati, A. Kumar, I. M. Held, and R. Gudgel,  
4 2011: Statistical-dynamical predictions of seasonal North Atlantic hurricane  
5 activity, *Monthly Weather Review*, **139**(4), 1070-1082.
- 6 —, R. Msadek, W. Anderson, Y. C. Chang, T. Delworth, K. Dixon, R. Gudgel, A.  
7 Rosati, B. Stern, , G. Villarini, A. Wittenberg, X. Yiang, R. Zhang, S. Zhang, and  
8 F. Zeng, 2012: Multi-year predictions of North Atlantic hurricane frequency:  
9 Promise and limitations, *Journal of Climate* (submitted).
- 10 Villarini, G., and G.A. Vecchi, 2012. North Atlantic Power Dissipation Index (PDI) and  
11 Accumulated Cyclone Energy (ACE): Statistical modeling and sensitivity to sea  
12 surface temperature changes. *Journal of Climate* **25**(2), 625-637.
- 13 —, —, and J.A. Smith, 2010: Modeling of the dependence of tropical storm counts  
14 in the North Atlantic Basin on climate indices. *Monthly Weather Review* **138**(7),  
15 2681–2705.
- 16 —, —, and —, 2012: U.S. landfalling and North Atlantic hurricanes: Statistical  
17 modeling of their frequencies and ratios. *Monthly Weather Review*, **140**, 44–65,.
- 18 —, —, T.R. Knutson, M. Zhao and J.A. Smith, 2011a: Reconciling Differing Model  
19 Projections of Changes in the Frequency of Tropical Storms in the North Atlantic  
20 Basin in a Warmer Climate, *J. Climate*, **24**(13), 3224-3238, doi:  
21 10.1175/2011JCLI3853.1.

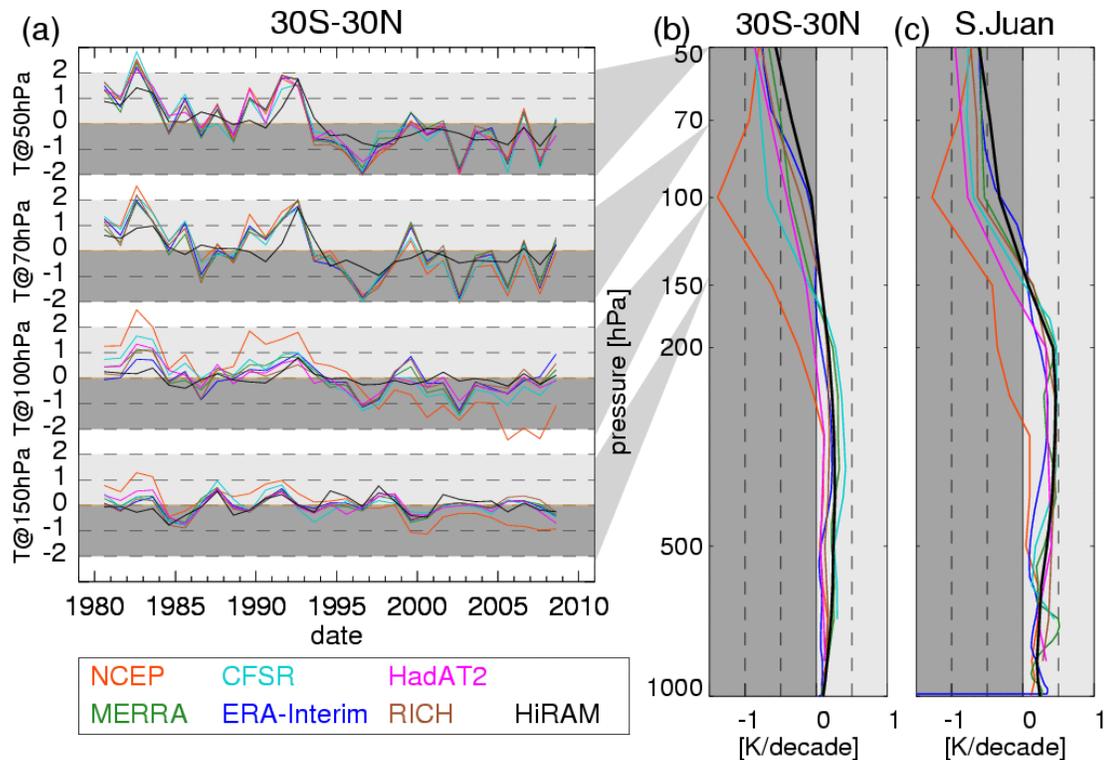
- 1 ———, ———, ———, and J.A. Smith, 2011b: Is the Recorded Increase in Short Duration  
2 North Atlantic Tropical Storms Spurious? *J. Geophys. Res.*  
3 doi:10.1029/2010JD015493
- 4 WMO (World Meteorological Organization), 2011: Scientific Assessment of Ozone  
5 Depletion: 2010, GLObal Ozone Research and Monitoring Project -- Report No.  
6 52, 516 pp., Geneva, Switzerland.
- 7 Xie, S., C. Deser, G. A. Vecchi, J. Ma, H. Teng, and A. T. Wittenberg, 2010: Global  
8 warming pattern formation: Sea surface temperature and rainfall. *J. Climate*, **23**,  
9 966–986.
- 10 Yoshimura, J. and M. Sugi, 2005: Tropical cyclone climatology in a high-resolution  
11 AGCM – Impacts of SST warming and CO2 increase. *SOLA*, **1**, 133-136,  
12 doi:10.2151/sola.2005-035.
- 13 Zhao, M., and I. M. Held, 2012: TC-permitting GCM simulations of hurricane frequency  
14 response to sea surface temperature anomalies projected for the late 21st century.  
15 *J. Climate*, **25**(8), doi:10.1175/JCLI-D-11-00313.1.
- 16 ———, ———, S.-J. Lin, and G. A. Vecchi, 2009: Simulations of global hurricane  
17 climatology, interannual variability, and response to global warming using a 50-  
18 km resolution GCM. *J. Climate*, **22**, 6653–6678.
- 19 ———, ———, and G. A. Vecchi, 2010: Retrospective forecasts of the hurricane season  
20 using a global atmospheric model assuming persistence of SST anomalies.  
21 *Mon. Wea. Rev.*, **138**, 3858–3868.
- 22

1  
2  
3  
4  
5  
6

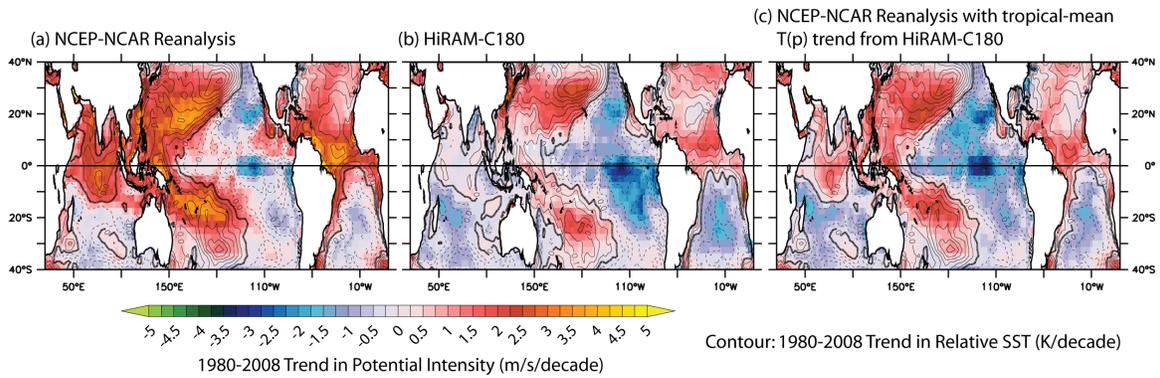
**Table 1:** Perturbation experiments run with HiRAM-C180 to explore impact of atmospheric cooling on TC frequency.

<b>Experiment Name</b>	<b>SST Forcing</b>	<b>Radiative Forcing</b>	<b>Cooling Location</b>	<b>Cooling Rate</b>
AMIP	Monthly 1979-2008 HadISST	Time-varying CO <sub>2</sub> , O <sub>3</sub> , natural and anthropogenic aerosols	--NA--	--NA--
CTL	Monthly HadISST climatology	Climatological CO <sub>2</sub> , O <sub>3</sub> and aerosols	--NA--	--NA--
B4	“ “	“ “	75-150hPa	0.25 K·day <sup>-1</sup>
B5	“ “	“ “	75-150hPa	0.5 K·day <sup>-1</sup>
C1	“ “	“ “	150-300hPa	0.5 K·day <sup>-1</sup>
C2	“ “	“ “	150-300hPa	2.0 K·day <sup>-1</sup>

7



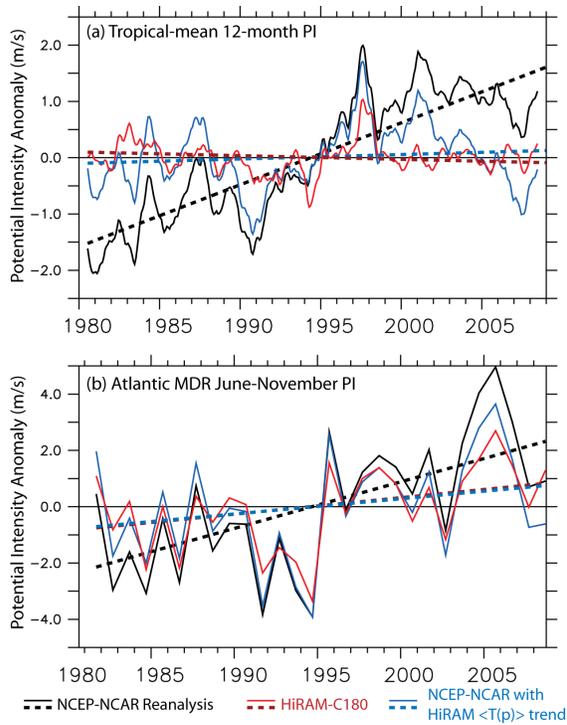
1  
 2 **Figure 1:** Estimates of 1980-2008 atmospheric temperature changes from six  
 3 observationally-based products, including four assimilation products (NCEP/NCAR  
 4 Reanalysis in red; NASA-MERRA in green, NOAA-CFSR in cyan, ECMWF-ERA-  
 5 Interim in dark blue) and two radiosonde-only products (UKMO-HadAT2 in pink and the  
 6 32-member ensemble-mean RICH in brown), and from a three-member ensemble mean  
 7 of the HiRAM-C180 AGCM. The left and center panels focus on tropical (30°S-30°N  
 8 averages) and the right panel shows the data point nearest the radiosonde station at San  
 9 Juan, Puerto Rico. Left time-series shows the evolution of annual-mean atmospheric  
 10 temperature anomalies at three levels in the TTL (150hPa-70hPa) and one in the lower  
 11 stratosphere. The center and right profiles show the linear least-squares trend of  
 12 temperature over 1980-2008 for each product. Values in the time-series are in K, values  
 13 of the trends are in K per decade.



1

2 **Figure 2:** 1980-2008 linear-least-squares trends in monthly Bister and Emanuel (1998,  
 3 2002) PI (shaded) and relative SST (contour) computed from (a) NCEP-NCAR  
 4 Reanalysis, (b) C180-HiRAM, and (c) adjusted NCEP-NCAR Reanalysis, in which the  
 5 tropical-mean air temperature trend is replaced with that from C180-HiRAM (see Eq. 1).  
 6 Relative SST is the difference between SST at a location and the tropical average (30°S-  
 7 30°N), with units of K.

8



1

2

3 **Figure 3:** Time-series of Bister and Emanuel (1989) PI from NCEP (black), HiRAM-

4 C180 (red) and the modified NCEP Reanalysis (blue), in which the tropical-mean

5 atmospheric temperature trend is replaced with that from HiRAM-C180. Panel (a)

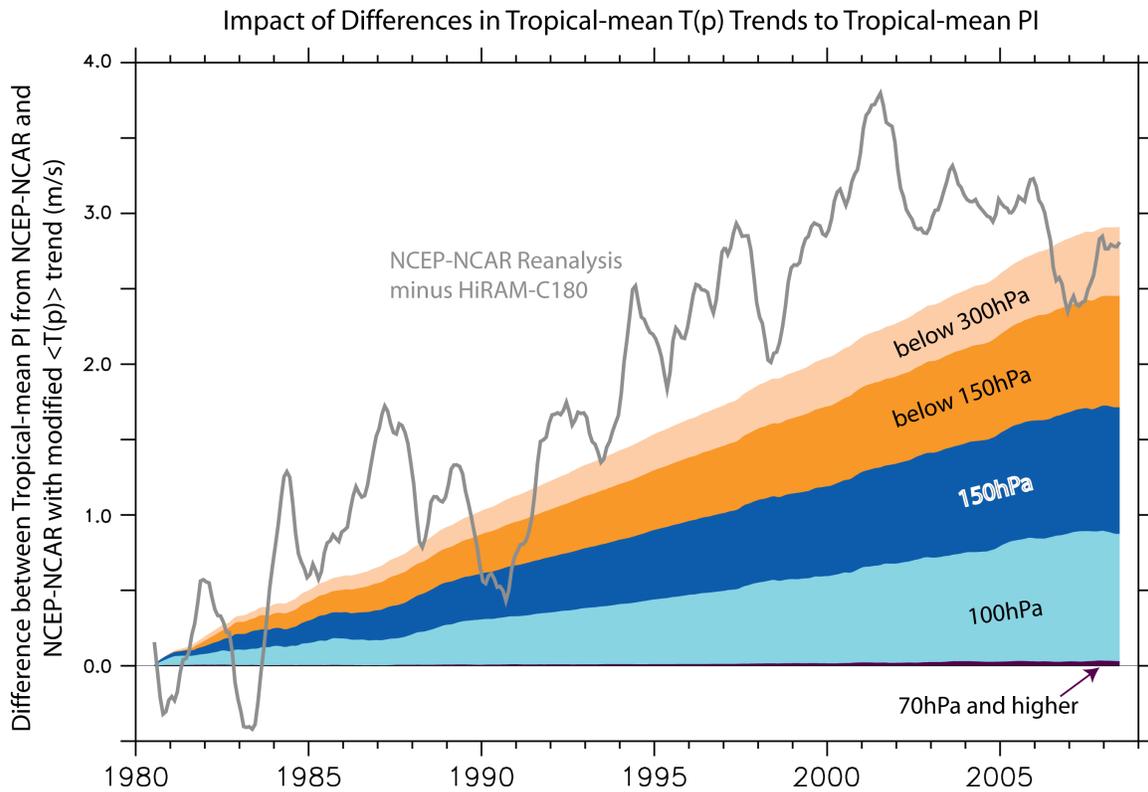
6 shows the twelve-month running mean of tropical-mean ( $0^{\circ}$ - $360^{\circ}$ ,  $30^{\circ}$ S- $30^{\circ}$ N) PI;

7 panel (b) shows the June-November average over the Atlantic hurricane main-

8 development region (MDR;  $80^{\circ}$ W- $20^{\circ}$ W,  $10^{\circ}$ N- $25^{\circ}$ N). Dashed lines show the

9 1980-2008 linear least-squares trends.

10

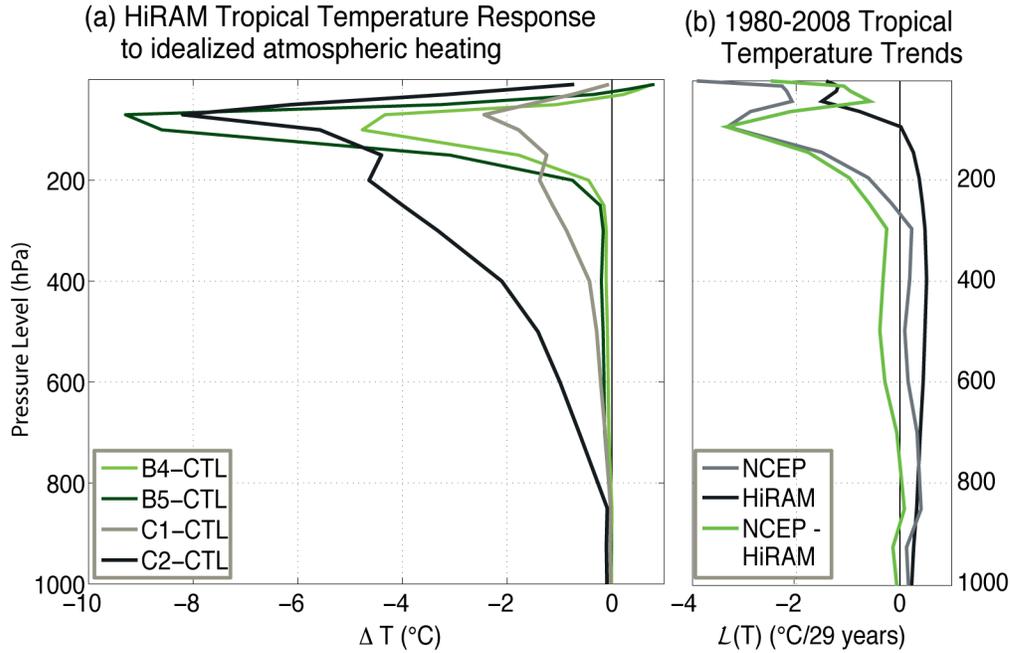


1

2

3 **Figure 4:** Contribution to PI differences between NCEP and HiRAM from different  
 4 trends in tropical temperature at different atmospheric levels. The gray line shows the  
 5 twelve-month running mean of the difference between tropical-mean PI in NCEP and  
 6 HiRAM. The different wedges show the impact on tropical-mean PI of replacing the  
 7 tropical-mean temperature trend in NCEP with that from HiRAM (see Section 4, Eq. 1).

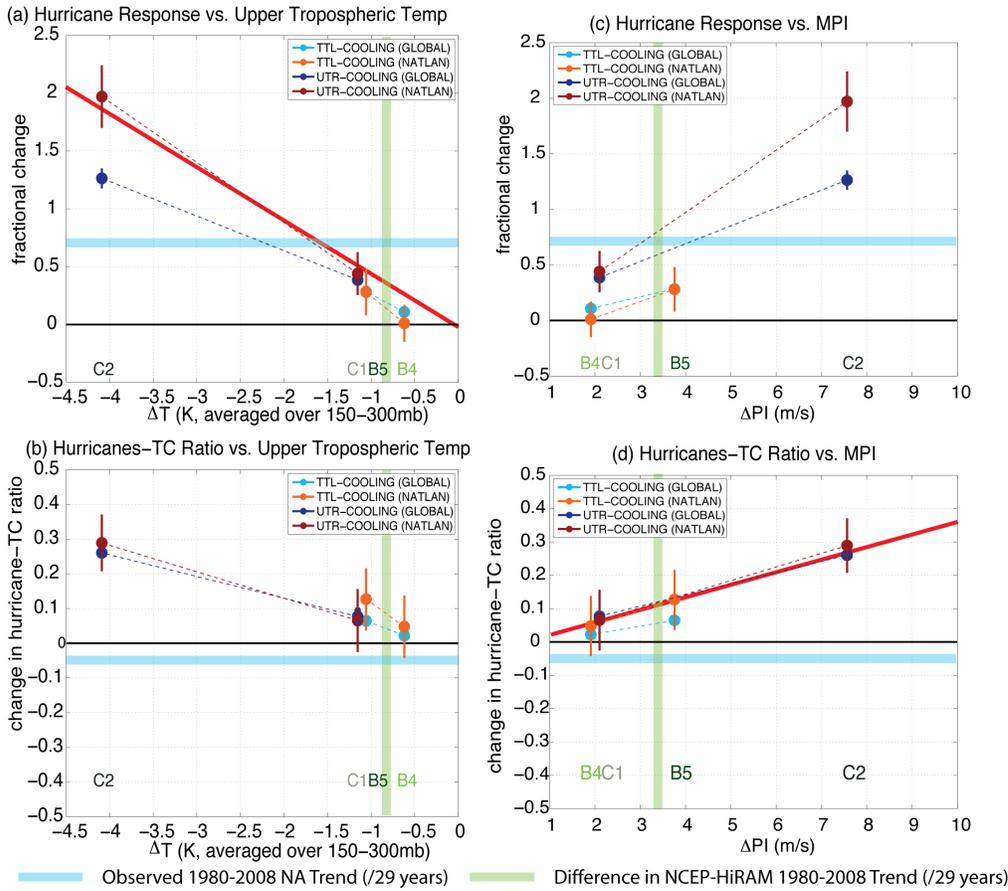
1



2

3 **Figure 5:** (a) Annual-mean tropical (30°S-30°N) atmospheric temperature response in  
 4 HiRAM-C180 to the idealized tropospheric and TTL heating anomalies described in  
 5 Table 1 and Section 5. (b) Tropical atmospheric temperature trends over 1980-2008 in  
 6 NCEP (gray) and HiRAM AMIP (black); green line shows the difference between the  
 7 NCEP and HiRAM AMIP trends.

8



1      ▬ Observed 1980-2008 NA Trend (/29 years)      ▬ Difference in NCEP-HiRAM 1980-2008 Trend (/29 years)

2      **Figure 6:** Response of TC activity in the Atlantic (red symbols) and globe (blue symbols)

3      from HiRAM-C180 to the tropospheric and TTL heating experiments described in Table

4      1. Left panels plot measures of TC activity against the tropical-mean change in upper-

5      tropospheric temperature. Right panels plot measures of TC activity against the tropical-

6      mean change in Bister and Emanuel (1998, 2002) PI. Shown are the fractional change in

7      hurricane frequency (top panels), and the change in the ratio of hurricane to TC

8      frequency (bottom panels). The blue horizontal bars indicate the observed 1980-2009

9      trends in North Atlantic activity based on HURDAT; the green vertical bars indicate the

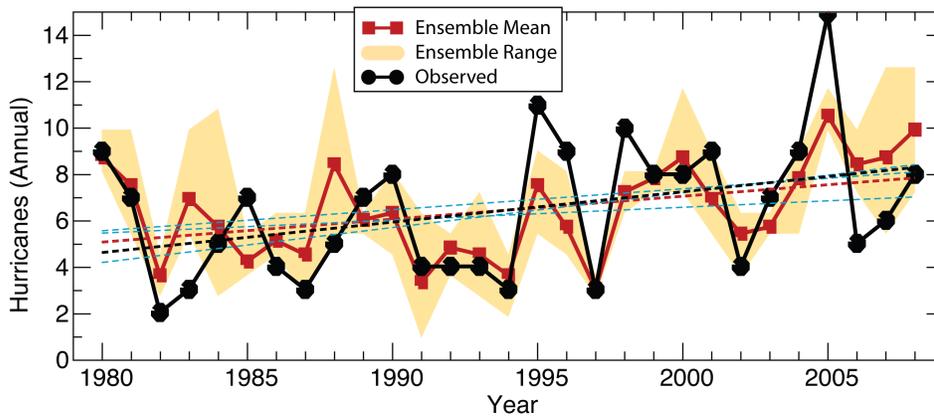
10      difference in the 1980-2009 trends between NCEP and HiRAM. In panels (a) and (d), the

11      solid diagonal red line shows the linear-least squares fit to the four North Atlantic points,

12      with a zero intercept.

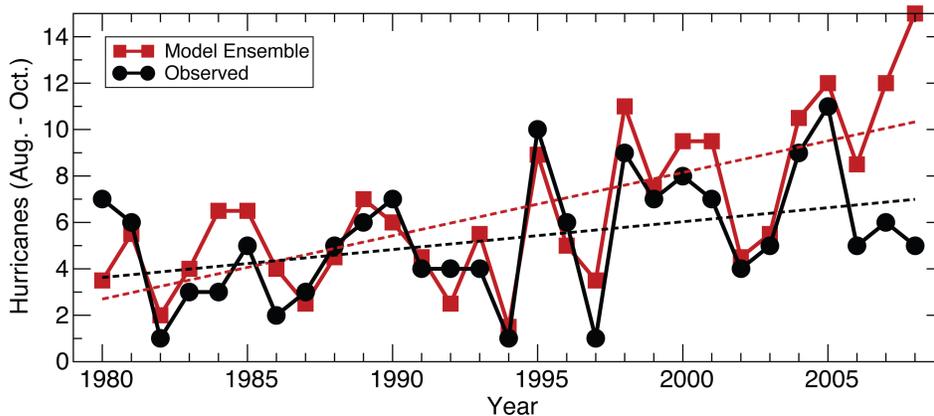
### Atlantic Hurricanes (1980-2008): HiRAM-Simulated vs. Observed

Correlation: ens-mean = 0.69; Linear trends: +0.10 storms/yr (model ens mean) +0.13 storms/yr (observed).  
 ens-range = [0.47,0.59] [+0.06, 0.15] storms/yr (model ens range)



### Atlantic Hurricanes (1980-2008): ZETAC-Simulated vs. Observed

Correlation = 0.69; Linear trends: +0.27 storms/yr (model) and +0.12 storms/yr (observed).



1  
 2 **Figure 7:** North Atlantic hurricane frequency in the HiRAM AGCM (Zhao *et al.* 2009,  
 3 2010; upper panel) and the ZETAC regional model (Knutson *et al.* 2007, 2008; lower  
 4 panel), compared against observations. Upper panel is for annual hurricane frequency,  
 5 lower panel is for August-October hurricane frequency. Black lines and circles show the  
 6 evolution of hurricane frequency in observations (dashed line shows linear trend). Thick  
 7 red line and squares shows the ensemble-mean evolution of each model; dark red dashed  
 8 line shows the trend of the ensemble mean. In the upper panel the light orange shading  
 9 shows the three member ensemble spread, and the dashed light blue lines show the trend  
 10 of each ensemble member.