1	Climatology and recent increase of westerly winds over
2	the Amundsen Sea derived from six reanalyses
3	
4	Thomas J. Bracegirdle
5	British Antarctic Survey, Cambridge, United Kingdom
6	
7	Prepared for the International Journal of Climatology
8	
9	Short title: Winds over the Amundsen Sea
10	
11	23 January 2012
12	
13	Full author address:
14	British Antarctic Survey, Cambridge, UK, CB3 0ET
15	e-mail: <u>tjbra@bas.ac.uk</u>
16	tel: +44 (0) 1223 221571
17	fax: +44 (0) 1223 362616
18	

#### 1 Abstract

2 The observed acceleration of glaciers from West Antarctica into the Amundsen Sea is estimated 3 to be contributing 6% to current sea-level rise with the estimated potential to add 0.24 m to 4 global sea level. Stronger westerly winds over the Amundsen Sea can increase the flow of 5 relatively warm ocean water to the base of ice shelves that flow from glaciers into the 6 Amundsen Sea. Thinning of the glaciers caused by this warming is a potentially important 7 factor in driving the observed acceleration of glaciers. However, the climatology of winds in the 8 region has not been extensively studied due to a lack of in-situ observational long-term records. 9 Here six different reanalysis datasets are assessed (CFSR, ERA-40, ERA-Interim, JRA-25, 10 MERRA and NNR1) to determine a best estimate of variability and change since 1979 when the 11 widespread monitoring of the atmosphere from satellites was introduced. A comparison with 12 independent mean sea level pressure (MSLP) data from ice drifting buoys shows that ERA-13 Interim is clearly the most accurate at capturing the details of individual weather systems over 14 the neighbouring Bellingshausen Sea, implying that it is also accurate over the Amundsen Sea. 15 In terms of climatological means, the five recently-produced (after ~2000) reanalysis datasets 16 show only small differences. Decadal variations of westerly winds congruent with the observed 17 increases in the southern annular mode (SAM) index are a consistent feature across the 18 reanalysis datasets. In particular, the strong seasonal dependence of observed trends in the SAM 19 (i.e. significant positive trends in the summer and autumn in recent decades) is also seen in the 20 strength of westerly winds over the Amundsen Sea. In terms of year-to-year variability, the 21 annual mean westerly winds over the Amundsen Sea were found to be significantly correlated 22 with the SAM in summer (r=0.35;  $p \le 0.05$ ) and ENSO in spring (September-November) (r=0.41; 23 p≤0.05).

24

### 1 1. Introduction

2 The Amundsen Sea Embayment (ASE) is a location where glaciers (principally the Pine Island 3 and Thwaites glaciers) are observed to have been thinning in recent decades (Shepherd et al., 4 2001). This thinning is associated with an acceleration of flow into the ocean (Rignot et al., 5 2008). The additional ice entering the ocean is estimated to be contributing approximately 6% to 6 current sea level rise (Shepherd and Wingham, 2007) with a potential total contribution of 7 approximately 0.24 m (Vaughan et al., 2006). Strong westerly winds over the Amundsen Sea 8 were shown in the modelling study of Thoma et al. (2008) to divert relatively warm ocean 9 currents onto the continental shelf (i.e. toward the ASE) below floating ice shelves at the base of 10 the glaciers. Melting of these ice shelves could reduce the latitudinal stress on the glacier and 11 explain at least some of the observed acceleration. However, atmospheric circulation over the 12 Amundsen Sea has not been extensively studied due to a lack of long-term in-situ observational 13 records. Recent work has focussed on temperatures over the adjacent West Antarctica (Ding et 14 al., 2011) and shown that tropical Pacific SST anomalies can affect atmospheric circulation over 15 the Amundsen Sea. However, a direct assessment of the westerly winds adjacent to the ACE has 16 not been conducted. The key question addressed here is whether increases in westerly winds 17 have occurred over the Amundsen Sea that might ultimately contribute to the acceleration of the 18 glaciers in the ASE.

19

The climatological mean winds over the Amundsen Sea exhibit a distinct seasonal cycle with more pronounced westerly winds during the winter and spring seasons (Thoma et al., 2008). This occurs in association with the movement of a climatological centre of low pressure, known as the Amundsen Sea Low, which moves zonally along ~70°S to be situated just east of the Amundsen Sea in summer and over Ross Sea in other months. In any given season there can be large departures from climatology, since the Amundsen Sea is a region of particularly large

mean sea-level pressure (MSLP) variability (Cullather et al., 1996; Lachlan-Cope and
 Connolley, 2003).

3

4 As a consequence of the lack of long-term in-situ observational data in the region the main 5 source of data for studying atmospheric circulation over the Amundsen Sea is reanalysis 6 datasets. Reanalyses compensate for the lack of in-situ data by using a wide range of 7 observational data to constrain a full-physics atmospheric model, which provides a best estimate 8 of the real full three-dimensional structure of the atmosphere through time. Satellite retrieved 9 soundings are the most important source of data in regions of sparse in-situ observations, since 10 they can be used to attain relatively accurate estimates of, for example, atmospheric 11 temperature. Here three new reanalyses are assessed (CFSR, ERA-Interim and MERRA) 12 alongside three more established reanalyses (ERA-40, JRA-25 and NNR1), see Table 1 for full 13 names and other details. MSLP data from buoys deployed in the Bellingshausen Sea in late 14 summer/autumn 2001 are used to assess their skill. King (2003) used this buoy dataset in an 15 assessment of ECMWF operational weather analyses since it is an independent dataset that is 16 not assimilated into analyses. 17 18 An important motivation for verifying the reanalyses is to determine the best choice of dataset 19 for forcing ocean models in this region. An advantage of assessing a large number of reanalyses 20 is that a measure of confidence in their output can be gained from the inter-model spread. 21 Agreement between the models implies that their output is not sensitive to their different model 22 structures and assimilation schemes. 23 24 This paper is structured as follows. In section 2 the reanalysis and drifter buoy datasets are 25 described. The results of the comparison between the reanalyses and buoy data are then

26 presented in section 3. In section 4 the climatology of the westerly geostrophic winds over the

27 Amundsen Sea is shown in detail followed by an assessment of trends and links with the

Southern Annular Mode (SAM) and El Nino-Southern Oscillation (ENSO) in section 5. This is
 followed by the discussion and conclusions in section 6.

3

#### 4 2. Data and methods

Six reanalysis datasets were considered in this assessment: CFSR, ERA-40, ERA-Interim, JRA25, MERRA and NNR1. Their full names and other details are shown in Table 1. Reanalyses are
atmospheric datasets constructed using a weather model to fill the gaps in space and time
between in-situ observations.

9

10 NNR1, ERA-40 and JRA-25 have been available for a number of years. Bromwich et al. (2007) 11 provide a comprehensive review of the performance of these more established reanalyses over 12 the polar regions. They exhibit large biases in the Southern Hemisphere troposphere before the 13 widespread introduction of satellite-retrieved soundings in 1979 (Hines et al. 2000; Marshall 14 and Harangozo, 2000; Renwick, 2004; Sterl, 2004). In the modern satellite era (i.e. since 1979) 15 NNR1 showed significant biases in mean MSLP around Antarctica persisting into the 1990s. 16 This was due to observations being rejected where differences from the model were large. Since 17 1979 a steady increase in the number of in-situ observations and upgrades/additions to satellite-18 borne remote sensing has resulted in an increase in the number of accepted observations. In 19 particular many Australian automatic weather stations were assimilated in the mid 1990s, which 20 was probably the main contributor in ending an MSLP bias in NNR1 over East Antarctica 21 (Hines et al., 2000; Marshall and Harangozo, 2000; Bromwich and Fogt, 2004). ERA-40 shows 22 consistently small biases in MSLP after 1979 when compared to in-situ observational records. 23 Bromwich et al. (2007) found better agreement between JRA-25 and ERA-40 than between 24 NNR1 and ERA-40, indicating that JRA-25 biases are also relatively small during the modern

satellite era. Differences in SLP between all three are relatively small over the Amundsen Sea
 region (Bromwich et al., 2007).

3

Anomalies in the stratosphere are now recognised as having an impact on surface atmospheric
circulation (Thompson et al., 2005). Comparisons with observations in the stratosphere over the
poles are more difficult due to the lack of in-situ measurements. The most detailed assessments
are based on superpressure balloon campaigns such as those of *Manney et al.* (2005) and *Boccara et al.* (2008). As with the troposphere they find that ERA-40 is more accurate than
NNR1, but do not assess JRA-25.

10

11 In recent years a new generation of reanalyses have been launched (MERRA, ERA-Interim and 12 CFSR). Their skill in simulating Antarctic climate is not yet well known. Some of the important 13 aspirations/improvements are: (i) increased resolution, (ii) improved observational datasets, (iii) 14 more realistic representation of stratospheric dynamics (iv) improved assimilation and bias 15 correction of satellite radiances and (v) improved representations of the hydrological cycle. It is 16 notable that a 4D-Var data assimilation scheme is used in ERA-Interim, which would be 17 expected to incorporate the observational data in a more dynamically consistent way than the 18 3D-Var schemes used in the other reanalyses. More details can be found in the references listed 19 in Table 1. It should also be noted that MERRA and CFSR share common origins. In particular 20 both the input data and analysis system used for MERRA and CFSR are nearly the same (Saha 21 et al., 2010).

22

A dataset of MSLP measurements from drifting buoys deployed on sea ice in the

24 Bellingshausen Sea was used to compare the skill of the reanalysis datasets off coastal West

25 Antarctica. Since the Bellingshausen Sea is located only 500 km to the east of the Amundsen

- 26 Sea, this dataset should give a good indication of the performance of the reanalyses over the
- 27 Amundsen Sea. The MSLP dataset comprises data from three Metocean Compact Air-Launched

1	Ice Drifters (CALIBs) and intermittently spans the period 18 February to 21 May 2001. The
2	Global Telecommunications System (GTS) of the World Meteorological Organization (WMO)
3	did not have access to the data, which are therefore an independent source of information as
4	they were not assimilated into atmospheric analyses. From a deployment of three CALIBs on 18
5	February one failed shortly after but the other two (ARGOS IDs 21376 and 21388) were
6	successful. Another CALIB was deployed on 25 April (ARGOS ID 21392). However, King
7	(2003) points out that just after the deployment of CALIB 21392 three surface velocity profiling
8	barometers (SVPBs) were deployed approximately 135 km to the north and sent data to the
9	GTS. Therefore the approach of King (2003) is followed by focussing on data from CALIBs
10	21376 and 21388 over the period that they produced reliable data from 18 February through 15
11	March. For more details of the CALIB dataset including a map of buoy locations see King
12	(2003).
13	
14	For comparisons between reanalysis and observations, the CALIB observations made at the

model analysis times 00, 06, 12 and 18Z ± 1 hour were selected. A bilinear interpolation was
then used to extract MSLP from the gridded reanalyses at the CALIB locations. Where more
than one MSLP observation from a given CALIB occurs within 1 hour of one of the analysis
times, an average of these observations was used.

19

### **3. Comparison between reanalyses and drifting buoy**

#### 21 observations

22 Following King (2003) interpolated reanalysis data are compared to CALIBs 21376 and 21388

23 over the period 18 February through 15 March 2001 (Figure 1). From Figure 1 it is clear that

24 most of the reanalyses agree well with the CALIB data. However, relatively large differences of

25 ~5-10 hPa occur frequently for the JRA-25 and NNR1 reanalyses. JRA-25 also exhibits a

1	dependence on pressure, with larger positive biases for lower pressure, which indicates
2	particular problems with simulating cyclones. Such dependence is not evident for the other
3	reanalyses. The mean bias and variance (standard deviation) of the differences are plotted in
4	Figure 2a shows that in most cases the mean bias is within or very close to the CALIB
5	instrumental error of 1 hPa. The exceptions are the positive mean biases seen in NNR1
6	(CALIBs 21376 and 21388) and JRA-25 (CALIB 21388). The MSLP biases in NNR1 have
7	been well documented in previous studies (e.g. Bromwich and Fogt, 2004; Marshall, 2003).
8	However, the reason for particularly large differences between JRA-25 and 23188 is not clear.
9	However, comparing the different reanalyses over many years shows distinct year-to-year
10	variations in the relative biases over the Bellingshausen Sea (Figure 3). The slightly larger mean
11	bias in JRA-25 compared to the other reanalyses over the period spanned by the CALIB data
12	cannot be taken as representative of the full period since 1979. Indeed it is clear from Figure 3
13	that, apart from NNR1, there is generally good agreement between the reanalyses.
14	
15	The variance about the mean bias shows clearer differences between different reanalyses
16	(Figure 2). ERA-Interim shows the smallest variance, which indicates that it is the most
17	accurate at simulating specific weather events such as cyclones and fronts. ERA-40 shows the
18	next smallest variance, despite having a lower spatial resolution than MERRA and CFSR. This
10	
19	is possibly an indication that ERA-40 makes better use of available observations to constrain the
20	is possibly an indication that ERA-40 makes better use of available observations to constrain the model. NNR1 shows the largest variance . Only ERA-Interim matches the performance of the
20 21	is possibly an indication that ERA-40 makes better use of available observations to constrain the model. NNR1 shows the largest variance . Only ERA-Interim matches the performance of the ECMWF operational model that was the subject of analysis of King (2003).
20 21 22	is possibly an indication that ERA-40 makes better use of available observations to constrain the model. NNR1 shows the largest variance . Only ERA-Interim matches the performance of the ECMWF operational model that was the subject of analysis of King (2003).
<ol> <li>20</li> <li>21</li> <li>22</li> <li>23</li> </ol>	is possibly an indication that ERA-40 makes better use of available observations to constrain the model. NNR1 shows the largest variance . Only ERA-Interim matches the performance of the ECMWF operational model that was the subject of analysis of King (2003). For a short period CALIBs 21376 and 21388 were transmitting simultaneously. Therefore the
20 21 22 23 24	is possibly an indication that ERA-40 makes better use of available observations to constrain the model. NNR1 shows the largest variance . Only ERA-Interim matches the performance of the ECMWF operational model that was the subject of analysis of King (2003). For a short period CALIBs 21376 and 21388 were transmitting simultaneously. Therefore the MSLP gradient between them could be compared with the reanalyses. Figure 2b shows
<ol> <li>20</li> <li>21</li> <li>22</li> <li>23</li> <li>24</li> <li>25</li> </ol>	<ul> <li>is possibly an indication that ERA-40 makes better use of available observations to constrain the model. NNR1 shows the largest variance . Only ERA-Interim matches the performance of the ECMWF operational model that was the subject of analysis of King (2003).</li> <li>For a short period CALIBs 21376 and 21388 were transmitting simultaneously. Therefore the MSLP gradient between them could be compared with the reanalyses. Figure 2b shows differences between the gradients recorded by the CALIBs and those extracted from reanalysis</li> </ul>
<ol> <li>20</li> <li>21</li> <li>22</li> <li>23</li> <li>24</li> <li>25</li> <li>26</li> </ol>	is possibly an indication that ERA-40 makes better use of available observations to constrain the model. NNR1 shows the largest variance . Only ERA-Interim matches the performance of the ECMWF operational model that was the subject of analysis of King (2003). For a short period CALIBs 21376 and 21388 were transmitting simultaneously. Therefore the MSLP gradient between them could be compared with the reanalyses. Figure 2b shows differences between the gradients recorded by the CALIBs and those extracted from reanalysis data. It is difficult to draw robust conclusions from this, since the model-CALIB differences are

2	Results from just one or two months are of course unlikely to be fully representative of the
3	longer-term performance of the reanalysis datasets. For instance the skill can vary with season.
4	There may also be variations across models in their skill at simulating specific weather patterns
5	such as blocking, which can vary in prevalence over a one-month time scale.
6	
7	The relative differences between the reanalyses in MSLP estimates over the Amundsen Sea are
8	very similar to those seen over the Bellingshausen Sea (Figure 3). This shows that the general
9	good agreement between the CALIB data and reanalyses over the Bellinghausen Sea is a good
10	indication of skill over the Amundsen Sea.
11	

## 12 **4. Geostrophic wind climatology and variability**

13 There is a distinct seasonal variation of the MSLP pattern over the Amundsen Sea. During most 14 of the year a climatological centre of low pressure (the Amundsen Sea Low) is located towards 15 the Ross Sea. However, during summer the Amundsen Sea Low moves to the east towards the 16 Bellingshausen Sea. Thoma et al. (2008) show that over the Amundsen Sea this gives mainly 17 westerly geostrophic winds for most of the year, with a period of weak easterlies in summer. 18 They use a westerly geostrophic wind index to capture this variability, which is the difference 19 between the mean along two latitudinal arcs at 67.5°S and 72.5°S. Both arcs span the longitude 20 range 100°W to 125°W. The region over which this is defined in indicated in Figure 4a. Here 21 the index is denoted as  $\Delta p_{AS}$ . A positive index therefore indicates westerly geostrophic wind (a  $\Delta p_{\rm AS}$  of 1 hPa corresponds to a geostrophic westerly wind of ~2 ms<sup>-1</sup>). 22

23

The annual mean climatology of MSLP as estimated from ERA-Interim is shown in Figure 4a.
This shows that the long-term mean annual average gives very weak MSLP gradients over the

Amundsen Sea. ERA-Interim was chosen due to its greater accuracy compared to the CALIB
data. CFSR and MERRA also span the full period from 1979. Climatological differences
between ERA-Interim and CFSR over this period are generally small away from land, as
demonstrated by Figure 4b. However, the differences do indicate a slightly more positive Δp<sub>AS</sub>
in ERA-Interim compared to CFSR. Almost identical differences were found between ERAInterim and MERRA (not shown). Larger differences become apparent over West Antarctica,
but MSLP is not a valid quantity over high elevations.

8

9 The rather stagnant annual average MSLP gradient is not representative of the majority of 10 months, or indeed seasons. A time series showing  $\Delta p_{AS}$  from 1979 through 2009 filtered to 11 seasonal (3-month) time scales shows the clear intra-annual fluctuations between westerly and 12 easterly near-surface winds (Figure 5). It is the periods of strong positive  $\Delta p_{AS}$  that have been 13 found to coincide with intrusions of circumpolar deep water onto the continental shelf toward 14 the ASE (Thoma et al., 2008). The different reanalyses generally show similar results and it is 15 difficult to distinguish between them in Figure 5.

16

17 The inter-model differences are more apparent with a twelve-month low-pass filter applied to the  $\Delta p_{AS}$  time series (Figure 6). The differences between the reanalyses are largest during the 18 19 1980s and reduce during the 1990s, in agreement with the idea that additions and improvements 20 in the observations have gradually reduced biases over the modern satellite era (Bromwich et 21 al., 2007). The position of each model in the inter-model range remains broadly constant over 22 time. For instance, during most of the period JRA-25 is at or near the top of the inter-model 23 range and NNR1 is at or near the bottom. From year to year however the relative bias among the 24 models changes quite markedly. The changes in bias over time are important when considering 25 long-term trends. The reanalyses that span the full period clearly show more positive values of 26  $\Delta p_{AS}$  from the early 1990s onwards, with a large change at approximately 1992. It is unlikely

that this large change is an artefact of changes in observational input data. Previous studies have identified such artefacts in a number of reanalyses, but not for the year 1992 (Onogi et al., 2007; Sakamoto and Christy, 2009; Bromwich et al., 2011; Screen and Simmonds, 2011). In addition these studies show that in general reanalyses systems respond differently to changes in input data, further suggesting that the change in 1992 is not an artefact. The reliability of this increase in  $\Delta p_{AS}$  will be discussed in the context of the SAM in the next section.

7

#### 8 5. Recent trends and links with large-scale variability

9 In order to assess long-term changes over the Amundsen Sea it is important to determine links 10 with larger scale weather patterns. If robust links are identified then it can be possible to link 11 regional change to factors such as the ozone hole (Marshall et al., 2006). The possibility of 12 developing an ice-core proxy for  $\Delta p_{AS}$  is also an important motivation for assessing large-scale 13 links. Of the four most recent reanalyses, CFSR, ERA-Interim and MERRA were used for 14 assessing variability. The other, JRA-25, was not used since it produces mean biases larger than 15 instrumental error (Figure 2) and shows and a strong dependence between bias and pressure 16 (Figure 1).

17

18 Month-to-month variability in  $\Delta p_{AS}$  estimated from ERA-Interim is strongly positively 19 correlated with MSLP variability over the South Pacific sector of the Southern Ocean, with a 20 maximum of 0.7 at approximately 62°S, 120°W (Figure 7). Somewhat weaker negative 21 correlations of between -0.2 and -0.4 extend over all Antarctica. A ring of small positive 22 correlations at ~40°S indicates only a weak positive correlation with the SAM index. The results 23 from CFSR and MERRA data are almost identical (not shown). 24

1	Time series of annual mean $\Delta p_{AS}$ calculated from CFSR, ERA-Interim and MERRA are shown
2	in Figure 8 along with the SAM index of Marshall (2003), which is based entirely on long-term
3	in-situ observations. For the purposes of comparison with the SAM index the $\Delta p_{AS}$ time series
4	are normalised by their standard deviation. There is little correspondence between $\Delta p_{AS}$ and the
5	SAM index in terms of year-to-year variability (r=0.00 for ERA-Interim). However, a change
6	from lower SAM index values in the 1980s to higher values since the early 1990s broadly
7	correlates with the decadal changes in $\Delta p_{AS}$ seen in the reanalysis datasets (Figure 6).
8	
9	If the trends in the SAM are indeed linked to the trends in $\Delta p_{AS}$ then one would expect a strong
10	seasonal dependence in the trends in $\Delta p_{AS}$ that matches the well-documented seasonal
11	dependence of trends in the SAM (Thompson and Solomon, 2002). The summer (December-
12	February) and autumn (March-May) trends in $\Delta p_{AS}$ are much more strongly positive and follow
13	the decadal changes in the SAM (Figure 9). Both indices show a rapid summer and autumn
14	increase from the mid 1980s to ~2000 after which they return to more neutral values. The
15	correlations are only significant ( $p \le 0.05$ ) in summer. For winter there is no correlation
16	between the SAM and $\Delta p_{AS}$ (r=0.01 for both CFSR and MERRA and r=-0.00 for ERA-Int) and
17	neither index shows a detectable linear trend (Figure 9).
18	
19	In addition to correlations with the SAM, previous studies indicate relatively strong correlations
20	between ENSO and pressures over the Amundsen Sea of up to 0.5 in winter (e.g. Harangozo,
21	2000). The correlation between $\Delta p_{AS}$ estimated from ERA-Interim and the NINO3.4 index in
22	winter is 0.24, which is comparatively weak. A much stronger correlation coefficient of 0.41
23	(significant for $p \le 0.05$ ) is found for spring (Figure 10) and a much weaker correlation of 0.17
24	(insignificant for $p > 0.05$ ) in both summer and autumn. Calculations based on CFSR and
25	MERRA gives almost identical correlation coefficients.

#### **6. Discussion and conclusions**

In this paper the climatology and variability of westerly geostrophic wind over the Amundsen
Sea has been documented. This is potentially an important factor in affecting the melt rate at the
base of ice shelves that flow out from glaciers into the Amundsen Sea (Thoma et al., 2008).

5

6 There are no long-term in-situ observational records over the Amundsen Sea or the surrounding 7 region. Reanalysis datasets were therefore the main source of meteorological data used in this 8 study. Three new reanalyses (CFSR, ERA-Interim and MERRA) were assessed alongside three 9 more established reanalyses (ERA-40, JRA-25 and NNR1). A comparison with independent 10 mean sea level pressure (MSLP) data from ice drifting buoys (CALIBs) showed that ERA-11 Interim is clearly the most accurate over the neighbouring Bellingshausen Sea, implying that it 12 is also accurate over the Amundsen Sea. Other reanalyses were found to be less accurate (i.e. 13 larger bias variances), but in almost all cases showed mean biases within the CALIB 14 instrumental error. This implies that relative to ERA-Interim the other reanalyses are not as 15 successful at capturing the details of individual weather systems, but with the exception of 16 NNR1 produce a similarly accurate mean state. CFSR, ERA-Interim and MERRA were used for 17 the correlation analysis back to 1979. JRA-25 was not used since it showed a larger bias 18 variance than the other reanalyses and a strong dependence between bias and pressure. 19 20 Strong westerly winds over the Amundsen Sea are associated with a distinct high-pressure ridge 21 that is centred at the same longitude as the Amundsen Sea and just north of 60°S (Figure 7). 22 This is consistent with the case study of strong westerly winds conducted by Thoma et al.

23 (2008), which shows a high pressure ridge in the same location in spring 1994. The strong

24 correlation with pressures over the extreme South Pacific is probably the reason for the positive

25 correlation with ENSO (NINO3.4), which is strongest (0.41;  $p \le 0.05$ ) in the spring season.

26

1 Year-to-year correlations between  $\Delta p_{AS}$  and the SAM are much smaller, but trends in both 2 indices follow the same decadal pattern and seasonal dependence. This indicates that an external 3 underlying factor (or a number of factors) influences the decadal changes in both. Modelling 4 studies show strong evidence that the annual cycle in the SAM trends can be explained by the 5 formation of the ozone hole, the effect of which propagates down from the stratosphere to affect 6 the troposphere in winter and autumn (Thompson and Solomon, 2002). It is possible that in 7 addition to projecting on the SAM (i.e. increasing the strength of the circumpolar westerly 8 winds over the Southern Ocean), the westerly winds over the Amundsen Sea are also affected 9 by the ozone hole. The wider implication of the positive trends in  $\Delta p_{AS}$  is that there are ongoing 10 changes that act to increase the frequency/severity of intrusions of circumpolar deep water to the 11 inner continental shelf of the Amundsen Sea. Thoma et al. (2008) forced their ocean model 12 using NNR1 and found a decade of warming of shelf water in the 1990s. Our results show that 13 the increase in  $\Delta p_{AS}$ , to which this warming was attributed, is robustly reproduced by other 14 reanalyses. The close correspondence to decadal changes in the SAM further suggests that these 15 changes are not merely an artefact of the reanalysis models. 16

A number of questions remain for future work. How did  $\Delta p_{AS}$  vary before 1979 and can we use ice core proxies to determine this? To what extent have climate drivers such as the ozone hole, greenhouse gases and volcanic eruptions contributed to changes in  $\Delta p_{AS}$  and are any climate models sufficiently reliable in their simulation of the Amundsen Sea region to answer this question? Are the changes in  $\Delta p_{AS}$  documented here large enough to have a significant impact on ocean temperatures, hence on melting of ice shelves, hence on the flow speed of glaciers?

## 1 Acknowledgements

2 Thanks to Paul Holland and one anonymous reviewer for their comments, which greatly helped 3 to improve the manuscript. This study is part of the British Antarctic Survey Polar Science for 4 Planet Earth Programme. It was funded by UK The Natural Environment Research Council. The 5 European Centre for Medium Range Weather Forecasting are thanked for providing the ERA-40 and ERA-Interim datasets. The JRA-25 dataset used for this study was provided from the 6 7 cooperative research project of the JRA-25 long-term reanalysis by the Japan Meteorological 8 Agency (JMA) and the Central Research Institute of Electric Power Industry (CRIEPI). The 9 Global Modeling and Assimilation Office (GMAO) and the GES DISC are acknowledged for 10 the dissemination of the MERRA dataset. The CFSR data was retrieved from the Research Data 11 Archive, which is managed by the Data Support Section of the Computational and Information Systems 12 Laboratory at the National Center for Atmospheric Research in Boulder, Colorado. The NNR1 13 Reanalysis Derived data was provided by the NOAA/OAR/ESRL PSD, Boulder, Colorado, 14 USA, from their Web site at http://www.esrl.noaa.gov/psd/.

# 1 Bibliography

2	Boccara G, Hertzog A, Basdevant C, Vial F. 2008. Accuracy of NCEP/NCAR reanalyses and
3	ECMWF analyses in the lower stratosphere over Antarctica in 2005. Journal of
4	Geophysical Research-Atmospheres. 113(D20). 10.1029/2008jd010116.
5	Bromwich DH, Fogt RL. 2004. Strong Trends in the Skill of the ERA-40 and NCEP-NCAR
6	Reanalyses in the High and Midlatitudes of the Southern Hemisphere, 1958-2001.
7	Journal of Climate. 17(23): 4603-4619. 10.1175/3241.1.
8	Bromwich DH, Nicolas JP, Monaghan AJ. 2011. An Assessment of Precipitation Changes over
9	Antarctica and the Southern Ocean since 1989 in Contemporary Global Reanalyses.
10	Journal of Climate. 24(16): 4189-4209. 10.1175/2011jcli4074.1.
11	Bromwich DH, Fogt RL, Hodges KI, Walsh JE (2007), A tropospheric assessment of the ERA-
12	40, NCEP, and JRA-25 global reanalyses in the polar regions, Journal of Geophysical
13	Research-Atmospheres, 112(D10), D10111, doi: 10.1029/2006jd007859.
14	Cullather RI, Bromwich DH, Van Woert ML. 1996. Interannual variations in Antarctic
15	precipitation related to El Nino southern oscillation. Journal of Geophysical Research.
16	<b>101</b> : 19109-19118.
17	Dee DP, Uppala SM, Simmons AJ, Berrisford P, Poli P, Kobayashi S, Andrae U, Balmaseda
18	MA, Balsamo G, Bauer P, Bechtold P, Beljaars ACM, van de Berg L, Bidlot J,
19	Bormann N, Delsol C, Dragani R, Fuentes M, Geer AJ, Haimberger L, Healy SB,
20	Hersbach H, Holm EV, Isaksen L, Kallberg P, Kohler M, Matricardi M, McNally AP,
21	Monge-Sanz BM, Morcrette JJ, Park BK, Peubey C, de Rosnay P, Tavolato C, Thepaut
22	JN, Vitart F. 2011. The ERA-Interim reanalysis: configuration and performance of the

1	data assimilation system. Quarterly Journal of the Royal Meteorological Society.
2	<b>137</b> (656): 553-597. 10.1002/qj.828.
3	Ding Q, Steig EJ, Battisti DS, Kuttel M. 2011. Winter warming in West Antarctica caused by
4	central tropical Pacific warming. <i>Nature Geoscience</i> . <b>4</b> (6): 398-403. 10.1038/ngeo1129.
5	Harangozo SA. 2000. A search for ENSO teleconnections in the west Antarctic Peninsula
6	climate in Austral winter. International Journal of Climatology. 20: 663-679.
7	Hines KM, Bromwich DH, Marshall GJ. 2000. Artificial surface pressure trends in the
8	NCEP/NCAR reanalysis over the Southern Ocean and Antarctica. Journal of Climate.
9	<b>12</b> : 3940-3952. 10.1175/1520-0442(2000)013<3940:ASPTIT>2.0.CO;2.
10	Kalnay E, Kanamitsu M, Kistler R, Collins W, Deaven D, Gandin L, Iredell M, Saha S, White
11	G, Woollen J, Zhu Y, Chelliah M, Ebisuzaki W, Higgins W, Janowiak J, Mo KC,
12	Ropelewski C, Wang J, Leetmaa A, Reynolds R, Jenne R, Joseph D. 1996. The
13	NCEP/NCAR 40-year reanalysis project. Bulletin of the American Meteorological
14	Society. <b>77</b> : 437-471.
15	King JC. 2003. Validation of ECMWF sea level pressure analyses over the Bellingshausen Sea,
16	Antarctica. Weather and Forecasting. 18: 536-540.
17	Lachlan-Cope T, Connolley W. 2006. Teleconnections between the tropical pacific and the
18	amundsen-bellinghausens sea: Role of the El Nino Southern Oscillation. Journal of
19	Geophysical Research-Atmospheres. 111(D23). 10.1029/2005jd006386.
20	Manney GL, Allen DR, Kruger K, Naujokat B, Santee ML, Sabutis JL, Pawson S, Swinbank R,
21	Randall CE, Simmons AJ, Long C. 2005. Diagnostic comparison of meteorological
22	analyses during the 2002 antarctic winter. Monthly Weather Review. 133(5): 1261-1278.
23	10.1175/MWR2926.1.

1	Marshall GJ. 2003. Trends in the Southern Annular Mode from observations and reanalyses.
2	Journal of Climate. 16: 4134-4143.
3	Marshall GJ, Harangozo SA. 2000. An appraisal of NCEP/NCAR reanalysis MSLP viability for
4	climate studies in the South Pacific. Geophysical Research Letters. 27: 3057-3060.
5	Marshall GJ, Orr A, van Lipzig NPM, King JC. 2006. The impact of a changing Southern
6	Hemisphere Annular Mode on Antarctic Peninsula summer temperatures. Journal of
7	<i>Climate</i> . <b>19</b> (20): 5388-5404.
8	Onogi K, Tslttsui J, Koide H, Sakamoto M, Kobayashi S, Hatsushika H, Matsumoto T,
9	Yamazaki N, Kaalhori H, Takahashi K, Kadokura S, Wada K, Kato K, Oyama R, Ose
10	T, Mannoji N, Taira R. 2007. The JRA-25 reanalysis. Journal of the Meteorological
11	Society of Japan. 85(3): 369-432. 10.2151/jmsj.85.369.
12	Renwick JA. 2004. Trends in the Southern Hemisphere polar vortex in NCEP and ECMWF
13	reanalyses. Geophysical Research Letters. 31: L07209-L07207doi.
14	Rienecker MM, Suarez MJ, Gelaro R, Todling R, Bacmeister J, Liu E, Bosilovich MG,
15	Schubert SD, Takacs L, Kim G-K, Bloom S, Chen J, Collins D, Conaty A, Da Silva A,
16	Gu W, Joiner J, Koster RD, Lucchesi R, Molod A, Owens T, Pawson S, Pegion P,
17	Redder CR, Reichle R, Robertson FR, Ruddick AG, Sienkiewicz M, Woollen J. 2011.
18	MERRA: NASA's Modern-Era Retrospective Analysis for Research and Applications.
19	Journal of Climate. 24(14): 3624-3648. 10.1175/jcli-d-11-00015.1.
20	Rignot E, Bamber JL, Van Den Broeke MR, Davis C, Li YH, Van De Berg WJ, Van Meijgaard
21	E. 2008. Recent Antarctic ice mass loss from radar interferometry and regional climate
22	modelling. Nature Geoscience. 1(2): 106-110. 10.1038/ngeo102.

1	Saha S, Moorthi S, Pan HL, Wu XR, Wang JD, Nadiga S, Tripp P, Kistler R, Woollen J,
2	Behringer D, Liu HX, Stokes D, Grumbine R, Gayno G, Wang J, Hou YT, Chuang HY,
3	Juang HMH, Sela J, Iredell M, Treadon R, Kleist D, Van Delst P, Keyser D, Derber J,
4	Ek M, Meng J, Wei HL, Yang RQ, Lord S, Van den Dool H, Kumar A, Wang WQ,
5	Long C, Chelliah M, Xue Y, Huang BY, Schemm JK, Ebisuzaki W, Lin R, Xie PP,
6	Chen MY, Zhou ST, Higgins W, Zou CZ, Liu QH, Chen Y, Han Y, Cucurull L,
7	Reynolds RW, Rutledge G, Goldberg M. 2010. The NCEP Climate Forecast System
8	Reanalysis. Bulletin of the American Meteorological Society. 91(8): 1015-1057.
9	10.1175/2010bams3001.1.
10	Sakamoto M, Christy JR. 2009. The Influences of TOVS Radiance Assimilation on
11	Temperature and Moisture Tendencies in JRA-25 and ERA-40. Journal of Atmospheric
12	and Oceanic Technology. 26(8): 1435-1455. 10.1175/2009jtecha1193.1.
13	Screen JA, Simmonds I. 2011. Erroneous Arctic Temperature Trends in the ERA-40 Reanalysis:
14	A Closer Look. Journal of Climate. 24(10): 2620-2627. 10.1175/2010jcli4054.1.
15	Shepherd A, Wingham D. 2007. Recent sea-level contributions of the Antarctic and Greenland
16	ice sheets. Science. <b>315</b> (5818): 1529-1532. 10.1126/science.1136776.
17	Shepherd A, Wingham DJ, Mansley JAD, Corr HFJ. 2001. Inland thinning of Pine Island
18	Glacier, West Antarctica. Science. 291(5505): 862-864.
19	Sterl A. 2004. On the (in)homogeneity of reanalysis products. Journal of Climate. 17(19): 3866-
20	3873.
21	Thoma M, Jenkins A, Holland D, Jacobs S. 2008. Modelling Circumpolar Deep Water
22	intrusions on the Amundsen Sea continental shelf, Antarctica. Geophysical Research
23	Letters. 35(18). 10.1029/2008gl034939.

1	Thompson DWJ, Solomon S (2002), Interpretation of recent Southern Hemisphere climate
2	change, Science, 296, 895-899, doi: 10.1126/science.1069270.
3	Thompson DWJ, Baldwin MP, Solomon S. 2005. Stratosphere-troposphere coupling in the
4	Southern Hemisphere. Journal of the Atmospheric Sciences. 62(3): 708-715.
5	Uppala SM, Kallberg PW, Simmons AJ, Andrae U, Bechtold VD, Fiorino M, Gibson JK,
6	Haseler J, Hernandez A, Kelly GA, Li X, Onogi K, Saarinen S, Sokka N, Allan RP,
7	Andersson E, Arpe K, Balmaseda MA, Beljaars ACM, van de Berg L, Bidlot J,
8	Bormann N, Caires S, Chevallier F, Dethof A, Dragosavac M, Fisher M, Fuentes M,
9	Hagemann S, Holm E, Hoskins BJ, Isaksen L, Janssen PAEM, Jenne R, McNally AP,
10	Mahfouf JF, Morcrette JJ, Rayner NA, Saunders RW, Simon P, Sterl A, Trenberth KE,
11	Untch A, Vasiljevic D, Viterbo P, Woollen J (2005), The ERA-40 re-analysis,
12	Quarterly Journal of the Royal Meteorological Society, 131, 2961-3012, doi:
13	10.1256/qj.04.176.
14	Vaughan DG, Corr HFJ, Ferraccioli F, Frearson N, O'Hare A, Mach D, Holt JW, Blankenship
15	DD, Morse DL, Young DA. 2006. New boundary conditions for the West Antarctic Ice
16	sheet: Subglacial topography beneath Pine Island Glacier. Geophysical Research
17	Letters. 33(9). 10.1029/2005gl025588.
18	
19	
20	

# 1 Figure captions

2

Figure 1. CALIBs 21376 and 21388 compared to CFSR, ERA-40, ERA-Interim, JRA-25
MERRA and NNR1.

6	Figure 2. Differences between reanalyses (which are grouped as labelled along the abscissa)
7	and CALIB data. (left panel) For all CALIB data (90, 50 and 100 obs for CALIBs 21376, 21388
8	and 21392 respectively) and (right panel) for differences between simultaneous data points from
9	CALIB 21376 and 21388 (46 obs). The mean reanalysis minus CALIB differences are shown
10	by the symbol and the bars show the standard deviations. The different symbols in (a) denote
11	different CALIBs as indicated in the key. The dotted lines show the instrumental error. In (b) a
12	cumulative instrumental error was used since no information on the statistical distribution of
13	instrumental errors was available.
14	Figure 3. Sea-level pressure from reanalyses for (left) the Bellingshausen Sea (72.5S, 90W) and
15	(right) the Amundsen Sea (72.5S, 110W). A one-year low pass digital filter was applied to the
16	time series of monthly mean data.
17	
18	Figure 4. (a) ERA-Interim climatological annual mean MSLP for 1979-2009. (b) The difference
19	in climatological annual mean MSLP between ERA-Interim and CFSR for 1979-2009. The
20	region over which $\Delta p_{AS}$ is defined is indicated by the box in (a) and the location of the
21	Amundsen Sea Embayment is indicated by 'ASE'.
22	
23	Figure 5. Seasonal timescale variations of time series of the Amundsen Sea SLP index from the
24	
	various reanalysis datasets. The monthly data was smoothed using a low pass (3-month

3	various reanalysis datasets. The monthly data was smoothed using a low pass (12-month
4	threshold) digital filter.
5	
6	Figure 7. The spatial distribution of correlations between time series of monthly mean MSLP at
7	all grid points and the time series of monthly mean $\Delta p_{AS}$ . Prior to calculating correlations all
8	time series were processed as follows: (i) the multi-year mean for each month was subtracted,
9	(ii) each calendar month was normalised by the standard deviation of the year-to-year variability
10	for that month and (iii) a linear detrending was applied. The bold black line shows the contour
11	for a p-value of 0.05. Data from ERA-Interim.
12	
13	<b>Figure 8.</b> The Marshall SAM index (solid line) and normalised $\Delta p_{AS}$ (dashed line:CFSR,
14	dashed-dot line: MERRA, dashed-dot-dot-dot line: ERA-Interim). The time series of annual
15	mean $\Delta p_{AS}$ was normalised by the standard deviation of its year-to-year variability. The
16	correlation coefficient between the Marshall index and $\Delta p_{AS}$ from ERA-Interim (both linearly
17	detrended) is shown in the plot title.
18	
19	Figure 9. As in Figure 8 but for summer (DJF), autumn (MAM), winter (JJA) and spring
20	(SON). In each season the time series of $\Delta p_{AS}$ was normalised by the standard deviation of its
21	year-to-year variability in that season. For DJF the year refers to December. The correlation
22	coefficients between the Marshall index and $\Delta p_{AS}$ from ERA-Interim (both linearly detrended)
23	are shown in the plot titles.

Figure 6. Annual timescale variations of time series of the Amundsen Sea SLP index from the

2	<b>Figure 10.</b> As for Figure 8, but with $\Delta p_{AS}$ (dashed line) plotted with NINO3.4 (solid line) in
3	spring (SON). The correlation coefficient between the NINO3.4 index and $\Delta p_{AS}$ from ERA-
4	Interim (both linearly detrended) is shown in the plot title.

# 1 Tables

Name (full name)	Reference	Data period	Horizontal grid (approx. grid size at 50° latitude)	Model top (hPa)
CFSR (NCEP Climate	Saha et	1979-present	T382 (~ 34 km)	~ 0.27
Forecast System	al.(2010)			
Reanalysis)				
ERA-40 (European	Uppala et al.	September	T159 (N80; ~125	0.1
Centre for Medium	(2005)	1957-August	km)	
Range Weather		2002		
Forecasting 40 Year Re- analysis)				
ERA-Interim (European	Dee et	1979-present	T255 (N128; ~79	0.1
Centre for Medium	al.(2011)	-	km)	
Range Weather				
Forecasting Interim Re-				
analysis)				
JRA-25 (Japanese 25-	Onogi et al.	1979-present	T106 (~120 km)	0.4
year Reanalysis)	(2007)			
MERRA (NASA Modern	Rienecker et	1979-present	0.5°×0.67° (~50 km)	0.01
Era Retrospective-	al. (2011)			
Analysis for Research				
and Applications)				
NNR1 (National Centers	Kalnay et al.	1948-present	T62 (~210 km)	~ 3
for Environmental	(1996)			
Prediction (NCEP) /				
NCAR (National Center				
for Atmospheric				
Research) Reanalysis 1)				

Table 1. Reanalysis product details.



















