

1 Gulf Stream and Kuroshio Current are synchronized

2 **The atmospheric jet stream ties temperature variations of two distant ocean currents for the decadal time scale.**

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3 **Observational records show that sea surface temperatures along the Gulf Stream**
4 **and Kuroshio tend to synchronize at decadal time scales. This synchroniza-**
5 **tion, which we refer to as the Boundary Current Synchronization (BCS), is**
6 **reproduced in global climate models with high spatial resolution. Both in**
7 **observations and model simulations, BCS is associated with meridional mi-**
8 **grations of the atmospheric jet stream. Changes in the strength and path of**
9 **the ocean currents driven by the jet shifts lead to the synchronicity of surface**
10 **temperatures. Numerical simulations using a conceptual model and an atmo-**
11 **spheric general circulation model are consistent with a notion that BCS is an**
12 **interbasin air-sea coupled mode. The abnormally hot summer in 2018 over the**
13 **Northern Hemispheric extratropics is explained by the positive phase of BCS.**

14 The two warm ocean currents, the Gulf Stream and the Kuroshio, are located in the west-
15 ern boundaries of the Atlantic and the Pacific Oceans, respectively, so they are referred to as
16 the western boundary currents (WBCs) (1–3). Meanderings of WBCs and the associated sea
17 surface temperature (SST) variations have long been known to affect local weather and cli-
18 mate in the coastal metropolitan areas, mainly because WBCs transport heat from the tropics
19 to the extratropics and modulate cyclogenesis (4) and low cloud formation (5). More recently,
20 high-resolution satellite observations helped reveal that heat released from WBCs have pro-
21 found impacts on the entire troposphere (6–8). The Gulf Stream and the Kuroshio are known
22 as centers of action in the midlatitude intrinsic variability (9–11), and also serve as surface
23 fingerprints of low-frequency natural climate variability (e.g., Atlantic Meridional Overturning
24 Circulation (12), Pacific Decadal Oscillation (PDO) (13)). Thus, understanding WBCs have
25 major implications for paleoclimatology (14), climate modeling (15), and disentangling natural
26 variability from the anthropogenic climate change (16).

27 Although both WBCs have experienced significant warming during recent decade (18), tight
28 linkages between the Gulf Stream and the Kuroshio have never been discovered. More than a
29 decade ago, a monograph by Kelly and Dong (2004) (19) found a hint of the WBC covariability
30 in the upper ocean heat content data. They estimated that 26% of heat content variations over
31 the entire North Atlantic and Pacific were in phase. Nevertheless, because the data length and
32 the spatial resolution were limited at that time, it was difficult to detect a fine structure or
33 well-defined covariability between the two WBCs. Though some climatologists mentioned this
34 potential WBC covariability as an outstanding issue (2), their monograph is, to the best of our
35 knowledge, the only observational effort that was taken to explore a possible linkage between
36 the Gulf Stream and the Kuroshio.

37 In the present day, satellite-based high-resolution SST data records (20, 21) have become
38 long enough to begin thorough analyses in this vein. The SST distribution in early 2018 may

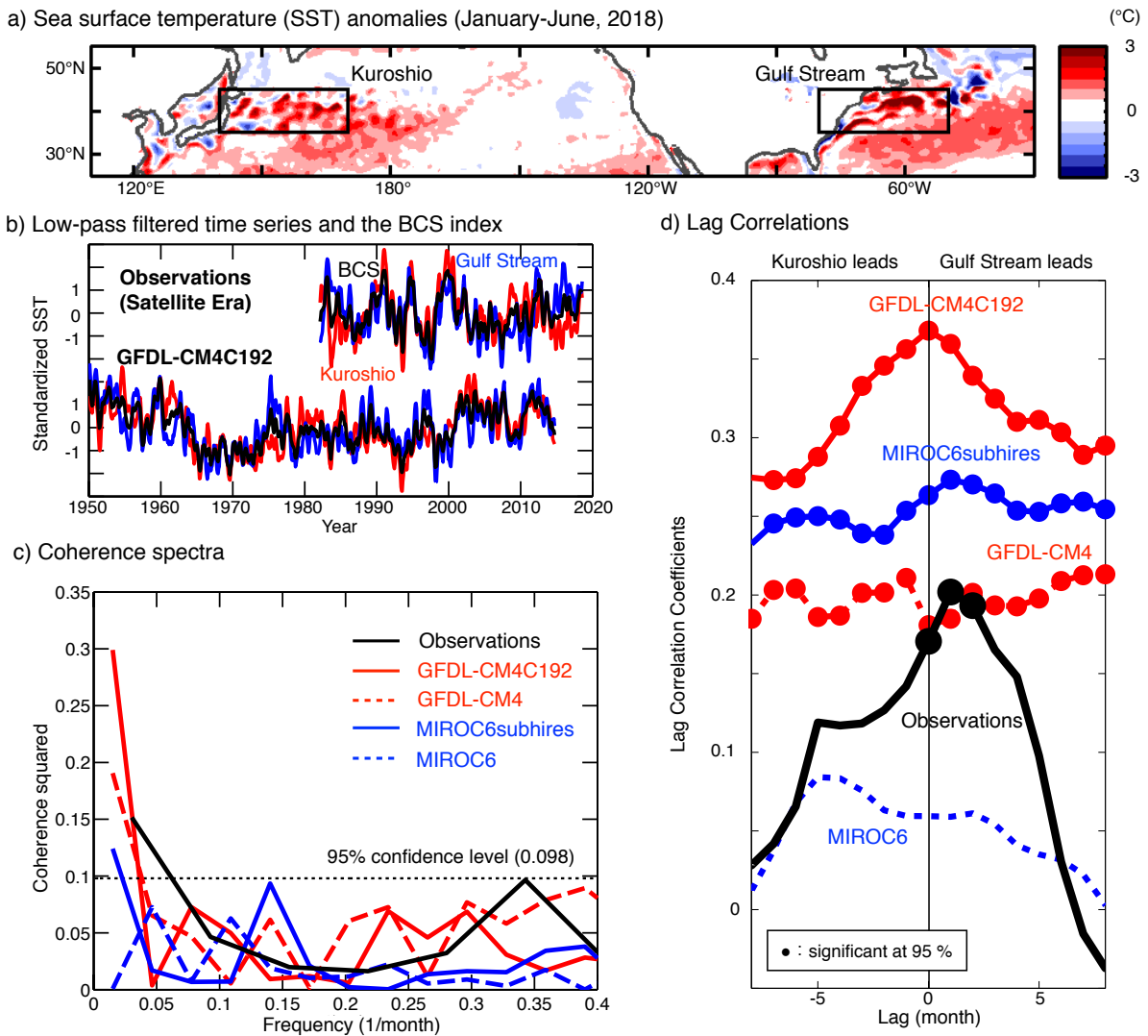


Figure 1: (a): Sea surface temperature (SST) anomalies averaged from January through June, 2018. Boxes show the locations of the two western boundary currents (WBCs). (b): Top, The observed BCS index (black) and five-month running-mean, standardized SST anomaly time series averaged over the Gulf Stream (35°N - 45°N , 80°W - 50°W) (blue dashed) and the Kuroshio (35°N - 45°N , 140°E - 170°E) (red dashed) regions defined as the boxes in (a). Bottom, As in top, but for a high-resolution global climate model (GCM). (c): Coherence spectra between the two SST anomaly time series averaged over the Gulf Stream and Kuroshio regions. The black line is observed data, the solid red and blue lines are from high-resolution GCMs, and the dashed lines are from low-resolution GCMs. The statistical significance calculated based on Amos and Koopmans (1963) (17) is shown as the black dotted line. (d): As in (c), but lag correlation coefficients. Positive lags means that Gulf Stream leads Kuroshio. Statistically significant correlations are shown as filled circles.

39 initiate speculations about a linkage between the two WBCs (Fig. 1a). During this time span,
40 SSTs near both of the WBC regions are warmer by about 3-5 °C than the temporal mean over
41 the past four decades, which corresponds to 2-3 standard deviations. This simultaneous warm
42 event would be rarely experienced by random chance. One could attribute these record-breaking
43 warm currents partly to the increasing greenhouse gas forcing, but this explanation appears not
44 to be the whole story as we shall see.

45 In this study, we show that the regional-mean SSTs over the two ocean currents are syn-
46 chronized for interannual to decadal time scales. First, the synchronization of the WBCs is
47 statistically demonstrated based on observed and modeled data analyses. Next, we define an
48 index to capture this covariability, as well as highlighting the impact of ocean resolutions on the
49 fidelity of simulated synchronization. Then, the physical mechanism is investigated based on
50 model experiments. Lastly, implications of this phenomenon are discussed.

51 **Statistical demonstration of the Boundary Current Synchronization (BCS)** Simple re-
52 gional mean SST time series are sufficient to suspect the existence of covariability between the
53 two warm currents. In Fig. 1b, based on satellite observations and output from a high-resolution
54 global climate model (GCM), GFDL-CM4C192, we plot five-month running-meaned, stan-
55 dardized time series of regional-mean SST anomalies over the Gulf Stream (35°N-45°N, 80°W-
56 50°W) and the Kuroshio (35°N-45°N, 140°E-170°E) regions. We hereafter investigate this
57 covariability by referring to it as the Boundary Current Synchronization (BCS).

58 Both in observations and high-resolution GCMs, the SSTs averaged over the two regions
59 exhibit significant coherence at the 95% confidence level only in frequencies lower than 0.05
60 /month (Fig. 1c). Here we plot the squared coherences for observations and four GCMs. GFDL-
61 CM4C192, which has finer atmospheric resolution than GFDL-CM4, exhibits higher coherence
62 at low frequency. MIROC6, which has lower oceanic resolution than MIROC6subhires, does

63 not exhibit a statistically significant coherence throughout all frequencies.

64 The SST variations are almost simultaneous between the two currents. In Fig.1d, we
65 plot lag correlations of the regional-mean SST anomalies between the two western boundary
66 current regions for observations and the four GCMs. As to observations, GFDL-CM4C192,
67 and MIROC6subhires, the highest correlations are realized within one-month lag between the
68 two regions, and they are both significant at the 95% confidence level. In observations and
69 MIROC6subhires, though Gulf Stream leads Kuroshio by one month, this small lag is not sig-
70 nificant considering the time scale of the phenomenon. By contrast, correlations in the low
71 resolution models are lower than observations and high resolution models (indeed, correlations
72 in MIROC6 are not significant at the 95% confidence level), and no strong peak at zero-lag is
73 detected.

74 **Definition of the BCS index and its dependence on spatial resolution** We define the BCS
75 index as the average of the low-pass filtered, standardized regional-mean SST anomalies over
76 the Gulf Stream (\tilde{G}) and the Kuroshio (\tilde{K}) regions, i.e., $BCS \equiv (\tilde{G} + \tilde{K})/2$ where a tilde
77 denotes performing a five-month running-mean filter and then normalizing by its own standard
78 deviation. As shown in Fig. 1b, the BCS index captures the temporal variations of SST over
79 both the Gulf Stream and the Kuroshio regions for both observations and models.

80 One might suspect that the regions used to define the index are too subjective and too large
81 to capture specific features of the boundary currents, and that they might reflect SST variability
82 in broader regions, rather than the currents themselves. Therefore, to present counterargument,
83 here we verify that the BCS index is almost equivalent to more objective and precise time series
84 that highlights variability of the boundary currents. In Fig. 2, we show the results from the
85 singular value decomposition (SVD) analysis between SST fields in the western North Pacific
86 and Atlantic regions. This analysis, which is also known as the maximum covariance analysis,

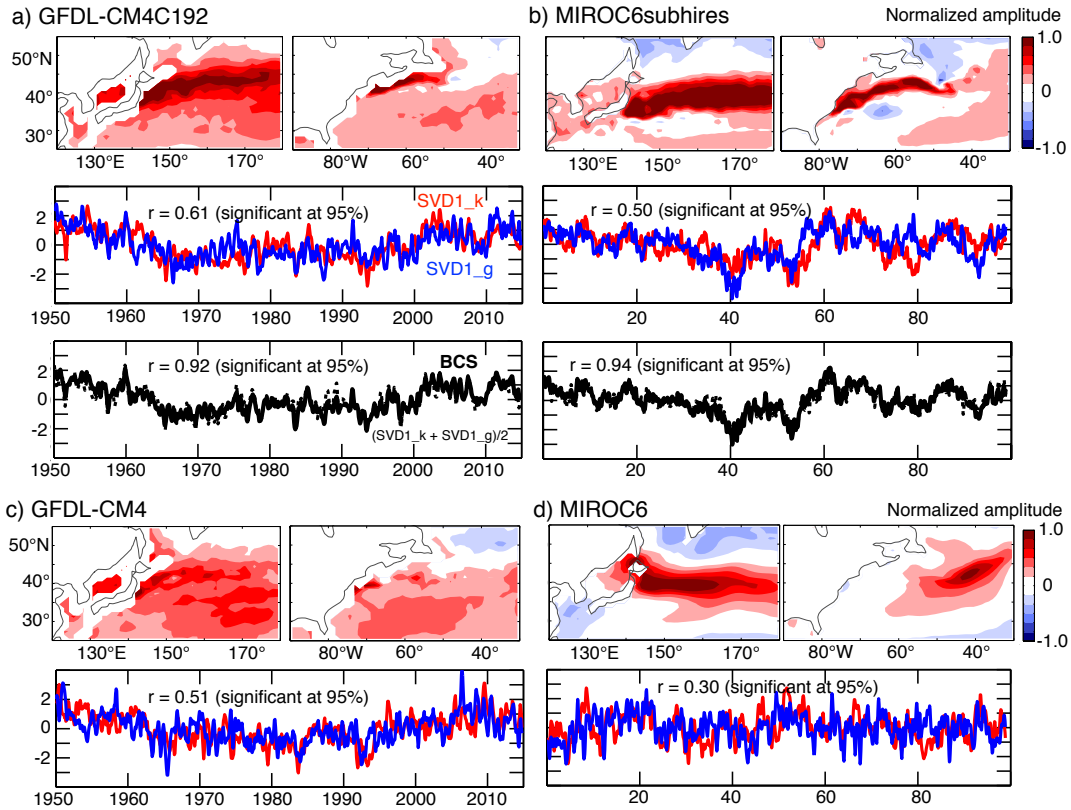


Figure 2: (a): Top, SST patterns of GFDL-CM4C192 extracted as the first mode of the singular value decomposition (SVD1) between the northwest Pacific (25°N-55°N, 120°E-180°) and the northwest Atlantic (25°N-55°N, 90°W-30°W) regions. Middle, Projected SST time series onto SVD1 for the northwest Pacific (SVD1_k) and the northwest Atlantic (SVD1_g). Three-month running-mean filtering is performed. Also shown are the correlation coefficient r between the two time series and its statistical significance. Bottom, As in middle, but the BCS index (solid) and the average of SVD1_k and SVD1_g (dashed). (b): As in (a), but for MIROC6subhires. (c): As in (a), but for GFDL-CM4. The bottom panel is omitted. (d): As in (c), but for MIROC6.

87 extracts the SST patterns that maximize the covariance between the two projected time series.

88 In two out of three models with eddy-permitting oceanic resolutions (i.e., GFDL-CM4C192
89 and MIROC6subhires), the first SVD (SVD1) mode captures SST variability of the narrow
90 boundary currents (Figs. 2a and 2b). The two projected time series exhibit statistically signif-
91 icant correlations of 0.61 and 0.50 for GFDL-CM4C192 and MIROC6subhires, respectively,
92 which confirms the suitability to perform SVD for these particular fields. Moreover, the mean
93 of the two projected time series exhibit a correlation larger than 0.9 with the BCS index. This
94 high correlation assures us that the simple definition of the BCS index is virtually identical to
95 a more objectively-defined index that reflects the temperature variations confined to the narrow
96 boundary current regions.

97 The SVD1 of GFDL-CM4, which also has an eddy-permitting ocean but has a coarser atmo-
98 sphere than GFDL-CM4C192, does not capture boundary currents well (Fig. 2c). Based on this
99 result, the low correlations shown in Fig. 1d is due to the ill-defined SST fronts that originate
100 from the coarse atmospheric resolution. In MIROC6, which does not resolve oceanic eddies,
101 the SVD1 does not capture the narrow boundary currents at all (Fig. 2d). Though the projected
102 time series exhibit significant correlations, the correlations are lower than the high-resolution
103 counterparts. This low correlations support a notion that high resolution models that adequately
104 resolve mesoscale eddies and their interactions with the atmosphere are essential for an accurate
105 representation of BCS. The BCS index represents boundary current variability only when the
106 spatial grids have sufficiently high resolutions in both atmospheric and oceanic components.

107 Though we have also performed the same analysis based on observations, SST variability is
108 captured clearly only in the Kuroshio, and that of the Gulf Stream is more subtle (Fig. S1, top).
109 This failure could be because the length of the data record over the satellite era is insufficient
110 for this particular analysis. Nevertheless, it is still notable that the SVD1 time series exhibit
111 statistically significant correlation of 0.49 (Fig. S1, bottom), which means that almost a quarter

112 variance of the first mode in these regions are explained by each other.

113 **Dynamical and thermodynamical manifestations of BCS** Meridional migrations of the tro-
114 pospheric westerly jet stream serve as an essential component of BCS. The regression map of
115 observed SST and zonal winds on the BCS index shows that, when the boundary current regions
116 are anomalously warm, the tropospheric jet stream tends to migrate northward, and vice versa
117 (Fig. 3a). The same relationship is reproduced by MIROC6subhires (Fig. 3b). Considering that
118 the two ocean currents are separated by the North American continent, they cannot exchange
119 heat via oceanic pathways within the decadal time scale. Therefore, it is virtually certain that
120 the atmospheric jet stream ties temperature variations of two distant ocean currents.

121 Wind-driven ocean dynamics, in addition to thermodynamical processes, is of first-order
122 importance for BCS. The composite maps of geostrophic current strength anomalies show that,
123 when the BCS index is positive, the positions for the boundary currents to be separated from
124 the shores tend to shift northward, and vice versa (Fig. 3c). For example, when BCS reaches +2
125 standard deviations, the Gulf Stream is separated from the North American continent near the
126 Virginia state (38° N). By contrast, when BCS reaches -2 standard deviations, the Gulf Stream
127 is diverted eastward near the South Carolina state (32° N). Similarly, the “ripping point” of the
128 Kuroshio also varies meridionally from the Iwate prefecture (40° N) to the Aogashima island
129 (32° N), depending on the phase of BCS.

130 This meridional shifts of the current pathways are consistent with the temperature variations
131 shown in Fig. 3b, considering that both the Gulf Stream and the Kuroshio transport warm water
132 from the tropics. A positive BCS event lets the boundary currents convey more heat to the north,
133 whereas a negative BCS event keeps heat to stay in the south.

134 Because BCS is associated with the low-frequency behavior of the atmospheric jet stream,
135 BCS has its implication for the midlatitude extreme weather. The spatial pattern of surface

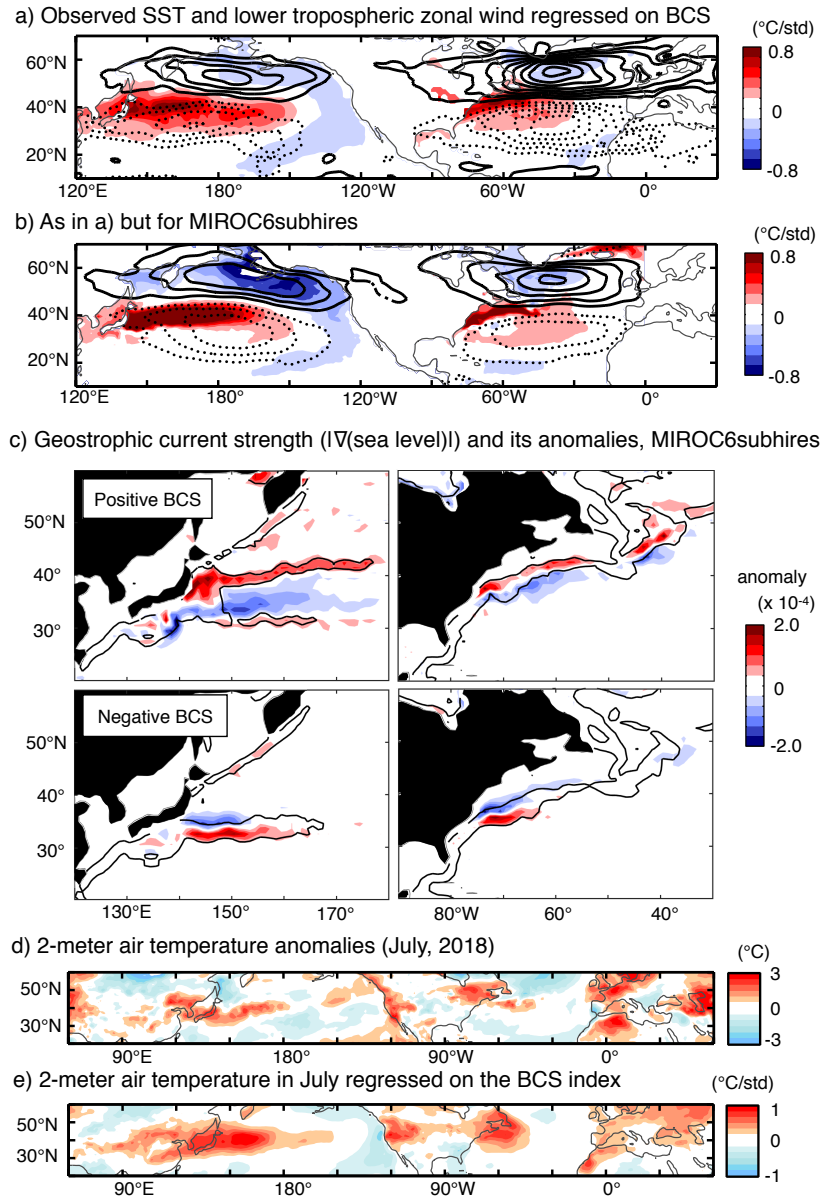


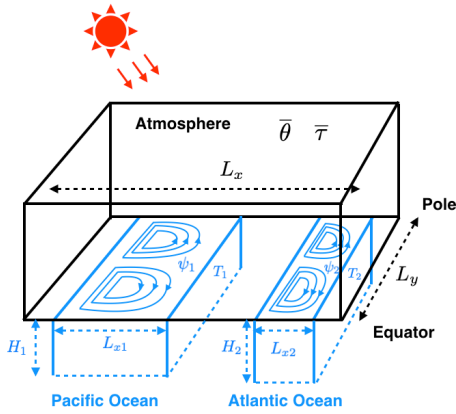
Figure 3: (a): Observed anomalies of SST (shaded areas) and zonal winds (contours) at 850 hPa regressed on the BCS index. Contour interval is 0.15 (m/s)/std. Solid (dashed) contours show positive (negative) anomalies, and zero contours are omitted. (b): As in (a), but for the MIROC6subhires model. (c): Composite maps of geostrophic current strength (contours) and its anomalies (shaded areas), which is estimated as anomalies of the absolute value of gradient of the sea level, for extremely positive (top) and negative (bottom) BCS events. Extreme BCS events are defined as the months when the BCS index exceeds its ± 2 standard deviations. Contour interval is 1.2×10^{-4} . (d): Monthly-mean surface temperature anomalies for July 2018. (e): 2-meter air temperature anomalies in July regressed on the BCS index calculated using the July-only data.

136 temperature anomalies observed in July 2018 (Fig. 3d) corresponds well to the regression map
137 of 2-meter air temperature anomalies on the BCS index calculated using the July-only data
138 (Fig. 3e). East Asia, the west and east coasts of North America, Europe, and Northwest Africa
139 experienced a hot summer in 2018, and these features are typical anomalies associated with
140 positive BCS events.

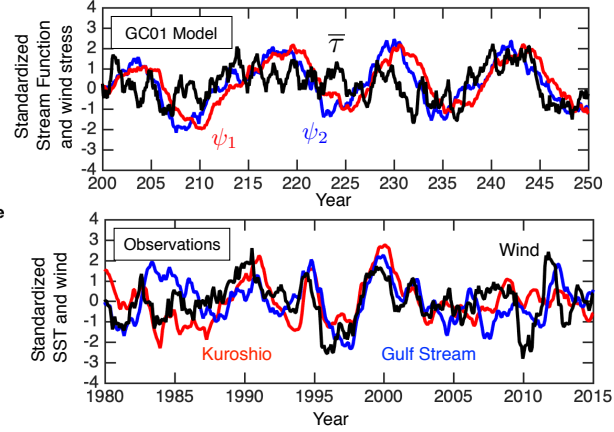
141 **Model experiments** Many previous studies convincingly showed possible physical processes
142 of interactions between large-scale tropospheric winds and WBCs (3, 7, 22–24). In particular, at
143 the beginning of this century, a possibly-related theoretical work was presented by Gallego and
144 Cessi (2001) (GC01) (25). They developed an idealized model of two ocean basins, each hav-
145 ing its own WBC, whose stream function is determined through the time-dependent Sverdrup-
146 balance (Fig. 4a). In their model, the two WBCs are coupled to each other only through the
147 zonally-symmetric atmosphere.

148 The idealized model by GC01 illustrates possible mechanisms of WBC covariability via
149 wind stress and heat fluxes, and theoretically predicts existence of a chaotic regime that exhibits
150 BCS-like variability. Using a modified version of the GC01 model with realistic choices of
151 parameters (see Data and Methods), the observed synchronicity of the two strengthening WBCs
152 and the westerly jet are reproduced (Fig. 4b). In particular, this conceptual model predicts that
153 the warm phase of WBCs are associated with a northward shift of jet stream and vice versa,
154 which is consistent with observations and the high-resolution GCMs. (Fig. 4c). As a promising
155 candidate to explain BCS, this model presents a physical process where zonally-symmetric
156 atmosphere synchronizes the Pacific and Atlantic Oceans with different intrinsic frequencies by
157 mediating the information of the two ocean currents. This inter-basin coupling mechanism is
158 also consistent with Omrani et al. (2019) (26), who showed, using semi-idealized atmospheric
159 general circulation models (AGCMs), that forcings from both Kuroshio and Gulf Stream are

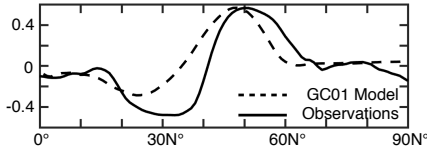
a) Schematic picture of the idealized model contracted by Gallego and Cessi (2001)



b) Modeled and observed time series of boundary currents and winds



c) Correlations between the BCS index and lower tropospheric zonal wind at each latitude



d) Regression map of lower tropospheric zonal wind on the BCS index calculated from the ensemble-mean of the MIROCsubhires AGCM runs

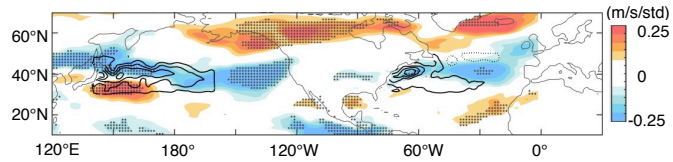


Figure 4: (a): Schematic diagram of the idealized model formulated by Gallego and Cessi (2001) (GC01). (b): Top, Modeled stream function anomalies ψ_1 (red) and ψ_2 (blue) at 4,000 km north of the southern boundary. Also shown in black is zonal-mean westerly wind stress ($\bar{\tau}$) anomalies at 5,200 km north of the equator. Bottom, Observed SST anomaly time series averaged over the Gulf Stream (35°N-45°N, 80°W-50°W) (blue) and the Kuroshio (35°N-45°N, 140°E-170°E) (red) regions. Also shown in black is observed zonal-mean zonal wind at 850 hPa averaged over the Northern Pacific and Atlantic (45°N-60°N, 140°E-0°). Each time series is normalized by its own standard deviation. (c): Solid, Observed correlation coefficients between the BCS index and 11-month running-mean zonal wind at 850 hPa zonally-averaged over the Pacific and the Atlantic (140°E-0°). Dashed, As in the solid curve, but between modeled temperature averaged over the two basin ($(\bar{T}_1 + \bar{T}_2)/2$) and the zonal wind stress ($\bar{\tau}$). The southern (northern) boundary of the model is set to be 20°N (90°N). (d): As in Fig. 3a, but for the ensemble-mean of the MIROCsubhires AGCM runs with SST anomalies added to climatology in the Kuroshio and Gulf Stream regions. Contour interval is 0.08 (m/s)/std.

160 necessary to maintain the zonally quasi-symmetric Northern Annular Mode variability.

161 To quantify the role of SST forcings, we have also performed an AGCM experiment. Figure
162 4d shows the regression map of zonal wind at 850 hPa on the BCS index calculated from the
163 ensemble-mean of the MIROCsubhires AGCM runs with observed SST anomalies added to the
164 model climatology only in the Kuroshio and Gulf Stream regions (“BCS experiment”; see Data
165 and Methods). This AGCM experiment is designed to capture a strongly positive BCS event. As
166 a forced response to warm western boundary current SST anomalies, the jet stream is shifted
167 northward. Similar results are obtained by subtracting zonal winds of the control experiment
168 from those of the BCS experiment.

169 These results qualitatively support the air-sea coupled mechanism proposed by GC01 in the
170 sense that zonally-symmetric atmospheric variability is forced by SST anomalies in the western
171 boundary regions, which in turn feeds back to the ocean through dynamical and thermodynamical
172 processes. In this regard, our additional sensitivity experiments using the GC01 model show
173 that, when the dynamical (thermodynamical) coupling is artificially removed, correlations be-
174 tween the Gulf Stream and the Kuroshio decline from 0.84 to 0.26 (0.41) (Fig. S2). This result
175 suggests that both dynamical and thermodynamical coupling mechanisms could play funda-
176 mental roles in BCS (see also Data and Methods and several previous studies (27–30)).

177 Quantitatively, however, zonal wind anomalies simulated by our AGCM experiment are
178 weaker than those of observations and the air-sea coupled model experiments. This small mag-
179 nitude of the anomalies can at least be explained by following two reasons. First, air-sea coupled
180 mechanisms, which are in principle not incorporated in an AGCM experiment, can amplify the
181 BCS variability. For example, air-sea couplings between the storm track and the northern hemi-
182 spheric WBCs (31) could enhance atmospheric variability to realize a realistic BCS amplitude.
183 Similarly, air-sea coupling processes could also operate for locating the SST front to an opti-
184 mum position where surface heating can enhance the persistence of the atmospheric intrinsic

185 variability (32). Second, as is also discussed in Smirmov et al. (2015) (33), the resolution of
186 observational SST data that are added to the lower boundary condition of the AGCM could
187 be too coarse to capture the full role of SST fronts. As shown in previous sections, GCMs
188 with high spatial resolution exhibit statistically significant BCS variability, whereas those with
189 low-resolution do not. This result is consistent with a notion that sharp SST fronts are another
190 essential ingredient of BCS, in addition to the inter-basin coupling mechanism proposed by
191 GC01. By adding idealized SST fronts to an AGCM experiment, Ogawa et al. (2012) (34) also
192 showed that the meridional position of the eddy-driven jet can be anchored by the SST fronts.

193 **Implications** Understanding BCS have immediate implications for human lives, because the
194 Gulf Stream and the Kuroshio transport heat from the tropics to the extratropics, and their
195 temperature variations affect the extreme weather of densely-populated areas in the northern
196 hemisphere (4–6). The hot summer experienced in 2018 is a good example of extreme weather
197 associated with BCS (Figs. 3d and 3e). In particular, because other prominent climate modes
198 (e.g., the El Niño Southern Oscillation) were relatively inactive in 2018, the BCS signature
199 may have clearly emerged in the observed air temperature over the entire northern hemispheric
200 extratropics.

201 BCS also have implications for fisheries productions because the variability of western
202 boundary currents modulates marine ecosystems (35–38). Warm SST associated with a north-
203 ward shift of the Gulf Stream increases the mortality of Atlantic cod (*Gadus morhua*) (35),
204 whereas migrations of pelagic fish, such as Japanese sardine (*Sardinops melanostictus*) (36)
205 and Pacific saury (*Cololabis saira*) (37, 38), are influenced by the Kuroshio variability because
206 they use the Kuroshio region as spawning and nursery grounds (39). Continuous monitoring of
207 the two WBCs with a finer observational network, as well as development of high resolution
208 GCMs are necessary for accurate understanding and prediction of BCS in a changing climate.

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271 **Data and Methods** Observed SST data is downloaded from the National Oceanic and Atmo-
272 spheric Administration (NOAA) Optimum Interpolation SST (OISST) (20) available at <https://www.esrl.noaa.gov/psd/data/gridded/data.noaa.oisst.v2.highres.html>.
273 Observational zonal wind fields at 850 hPa and 2-meter air temperature (July only) are
274 from the European Center for Medium range Weather Forecasting (ECMWF) ERA-Interim re-
275 analysis data (21) at <http://apps.ecmwf.int/datasets/data/interim-full-mode/levtype=sfc/>. The resolution of the OISST and ERA-Interim data sets used in this study
276 is 1° in both longitudes and latitudes. The time span used in this study is from December 1981
277 through September 2018.

280 The SST output of the GFDL models is from the Coupled Model Intercomparison Project
281 Phase 6 (CMIP6) data (40) available at the website of the World Climate Research Programme
282 (<https://esgf-node.llnl.gov/search/cmip6/>). The experiment considered in
283 this study is the first ensemble member of historical runs, which includes “historical” for CMIP6
284 for GFDL-CM4 and “hist-1950” (HighResMIP) for GFDL-CM4C192. GFDL-CM4 consists of
285 atmosphere and land models at about 100 km horizontal resolution and ocean and sea ice mod-
286 els at roughly 25 km horizontal resolution, whereas GFDL-CM4C192 has a higher atmospheric
287 spatial resolution (roughly 50 km resolution). For more detailed description of the GFDL mod-
288 els, see Held et al. (2019) (41). The time span used in this study is from January 1950 through
289 December 2014. For data analysis, the data is regridded via linear interpolation onto a 1° lon-
290 gitude by 1° latitude grid so it matches that of observational products.

291 In addition to the downloaded output from GFDL models, two kinds of state-of-the-art
292 climate models are used in the present study. One is the sixth version of Model for Interdis-
293 ciplinary Research on Climate (hereafter, MIROC6) which has been cooperatively developed
294 by a Japanese modeling community (42). The atmospheric and land surface components of
295 MIROC6 have horizontal resolution of a T85 spectral truncation. The model top of the at-
296 mospheric component is placed to 0.004 hPa, and there are 81 vertical levels. The governing
297 equations for the ocean and sea-ice components are discretized on tripolar horizontal coordi-
298 nate system with the resolution of nominal 1° , and there are 62 vertical levels. MIROC6 is
299 spun up for 1000 years with the preindustrial external forcing dataset following the protocol of
300 the sixth phase of the Coupled Model Intercomparison Project (CMIP6; (40)). After the model
301 climate reaches thermally and dynamically quasi-equilibrium state, additional 1000-year-long
302 integration is performed, and the last 100-year-long data of the preindustrial control simula-
303 tion is analyzed in the present study. The readers may refer to Tatebe et al. (2019) (42) for
304 detailed description and evaluation of MIROC6. The model data are distributed as Tatebe and
305 Watanabe (2018) (43) through the Earth System Grid Federation and are freely accessible. The
306 other is MIROC6subhires whose atmospheric and land surface components are the exactly same
307 as MIROC6 but the oceanic component is replaced by a horizontal higher-resolution version
308 with nominal 0.25° grid spacings in both of zonal and meridional directions. Because oceanic
309 mesoscale eddies and fronts are modestly resolved, a few parameterizations for subgrid hori-
310 zontal and isopycnal diffusion processes are set to be less effective in MIROC6subhires than
311 in MIROC6. MIROC6subhires is spun up for 700 years with initial conditions taken from the
312 preindustrial control simulation and the same external forcing dataset of MIROC6. After the
313 model climate reaches a quasi-equilibrium state, additional 200-year-long integration is per-
314 formed. The last 100-year-long data is analyzed in the present study.

315 To calculate detrended anomalies, we subtract monthly climatology (i.e., means of each

316 calendar month) and linear trends. The statistical significance of correlations is tested by the
317 two-tailed Student's t-test. To estimate statistical degrees of freedom in auto-correlated time
318 series, we employ a formula to calculate the effective sample size proposed by Bretherton et al.
319 (1999) (44).

320 We also conduct two AGCM experiments with 10 ensemble members using the atmospheric
321 component of MIROC6 and MIROC6Subhires. In the control experiment, the AGCM is forced
322 with the monthly climatology of SST and sea ice extent in MIROC6Subhires. Then, to examine
323 the impact of SST anomalies associated with BCS, the BCS experiment is conducted, where
324 observed interannual SST anomalies only in the Kuroshio (140°E-200°E, 30°N-50°N) and Gulf
325 Stream (80°W-20°W, 30°N-50°N) regions are superimposed on the SST climatology in the
326 control run. With different initial conditions, each experiment is integrated for five years from
327 1 January 1992, during which a large BCS index is observed, to investigate the response of the
328 zonal wind to SST anomalies associated with a strongly positive BCS event.

329 To describe the concept of BCS under a simple framework, we adopt a conceptual model
330 originally proposed by GC01 (25). As its detailed formulation and derivations of model equa-
331 tions have been already given by GC01, here we only provide a brief summary.

332 The model has two rectangular ocean basins coupled with a zonally periodic atmosphere,
333 as schematically illustrated in Fig. 4a. The atmosphere has zonal and meridional widths of L_x
334 and L_y , respectively, and each ocean basin (basin 1 and 2) has the same meridional extent as
335 the atmosphere. The zonal width of basin i ($i = 1, 2$) is $L_i = r_i L_x$, and its mean thermocline
336 depth is given by H_i . The dynamical and thermodynamical couplings between the atmosphere
337 and ocean are mediated by surface wind stress and air-sea heat flux, as detailed below.

338 The zonally periodic atmosphere is characterized by two variables, the zonally-averaged
339 surface potential temperature ($\bar{\theta}$) and zonal wind stress ($\bar{\tau}$). By considering conservations of
340 heat and zonal momentum as well as the quasi-geostrophic relation, we obtain the following

341 governing equations:

$$\bar{\theta} = \Gamma(\Lambda\bar{\theta}_A + r_1\lambda\bar{T}_{s1} + r_2\lambda\bar{T}_{s2} + Ff(y) - r_1\lambda\sigma_1\xi_1 - r_2\lambda\sigma_2\xi_2) + \sigma_3\xi_3 \quad (1)$$

342

$$\frac{\partial\bar{\tau}}{\partial y} = d_e\rho_o\nu_o\left\{\beta\left(y - \frac{L_y}{2}\right) + \frac{f_o}{Sd}(\bar{\theta} - \bar{\theta}_A)\right\} \quad (2)$$

343 with $\bar{\theta}_A = \frac{F_0 - A}{B}$, $f(y) = \cos\left(\frac{\pi y}{L_y}\right)$, $\Gamma = \frac{1}{C_{pa}\rho_o\nu_o d_e + B + (r_1 + r_2)\lambda}$, $\Lambda = C_{pa}\rho_o\nu_o d_e + B$,

344 and $d_e = \frac{Dd}{d + D}$. Here, \bar{T}_{s1} and \bar{T}_{s2} represent the zonally-averaged SST of basin 1 and 2,

345 respectively, and $\bar{\theta}_A$ is the planetary averaged potential temperature. Also, $\rho_a = \rho_o \exp(-z/D)$

346 is the atmospheric density (D is the scale height), $\nu = \nu_o \exp(-z/d)$ is the eddy relaxation

347 rate, and $F_0 + Ff(y)$ is prescribed shortwave radiation. In addition, C_{pa} denotes the specific

348 heat of the atmosphere, f_o is the Coriolis parameter, β is the planetary beta, S represents the

349 atmospheric stability, and λ is the damping coefficient of heat flux. For definition of other

350 parameters and their values, please see Table 1 and GC01. The zonal wind stress, $\bar{\tau}$, is obtained

351 by meridionally integrating Eq. (2) with the boundary condition that $\bar{\tau} = 0$ at $y = 0$.

352 To represent atmospheric variability, we introduce three stochastic forcings, $\sigma_1\xi_1$, $\sigma_2\xi_2$, and

353 $\sigma_3\xi_3$, with ξ_i ($i = 1, 2, 3$) denoting gaussian noise forcings with a zero mean and unit variance.

354 As described in Eqs. (1) and (2), the atmospheric state is determined by zonally-averaged SSTs

355 of the two basins.

356 The state variables of the ocean model are the zonally averaged SST ($=\bar{T}_{si}$) and interior

357 stream function at the western boundary ($=\psi_{Wi}$) of each basin ($i = 1, 2$). They are determined

358 by the upper ocean heat budget and linear baroclinic Rossby wave dynamics as follows (28):

$$\psi_{Wi}(y, t) = \frac{R_i^2}{\rho_w H_i} \int_{t - \frac{L_{xi}}{c_i}}^t \frac{\partial\bar{\tau}(y, t')}{\partial y} dt' \quad (3)$$

359

$$\frac{\partial\bar{T}_{si}}{\partial t} = \Upsilon_i \frac{\partial}{\partial y} (\psi_{Wi}^2 \frac{\partial\bar{T}_i}{\partial y}) - \frac{\lambda\Gamma}{C_{pw}\rho_w H_i} \{ \Lambda\bar{T}_i + \lambda r_j (\bar{T}_i - \bar{T}_j) - Ff(y) - \lambda r_i \sigma_i \xi_i - \lambda r_j \sigma_j \xi_j \} - \frac{\lambda\sigma_3\xi_3}{C_{pw}\rho_w H_i} + \epsilon \frac{\partial^2\bar{T}_i}{\partial y^2} \quad (4)$$

360 with $\Upsilon_i = \frac{C_{pw}\rho_w H_i}{2\lambda\delta_i L_{xi}}$. Here, $R_i (i = 1, 2)$ are the baroclinic radius of deformation of the basin i ,
 361 $c_i = \beta R_i^2$ denote the speed of the long Rossby waves, and δ_i are the zonal widths of frictional
 362 western boundary layers. Also, ρ_w is the seawater density, C_{pw} represents the specific heat of
 363 seawater, and ϵ is the horizontal diffusivity. Again, values and definitions of these variables
 364 are summarized in Table 1. Eq. (3) and (4) demonstrates that the ocean is dynamically forced
 365 by the atmosphere via the wind stress curl ($\frac{\partial \bar{\tau}}{\partial y}$; See Eq. (2)) and thermodynamical coupling
 366 is mediated by potential temperature ($\bar{\theta}$). Thus, the two-way coupling between the ocean and
 367 atmosphere is represented under a concise framework.

368 Using parameters shown in Table 1, we numerically integrate Eqs. (1)-(4) for 1,000 years
 369 with a time step of 0.5 month and discretized with a meridional grid spacing of 1×10^5 m. This
 370 experiment is referred to as the control (CTL) experiments. We have also performed calcula-
 371 tions with several different choice of parameters, and obtained similar BCS-like oscillations as
 372 demonstrated by GC01.

373 To understand the roles played by dynamical and themodynamical couplings, we perform
 374 two additional sensitivity experiments. In the "NoDYN" experiments, the model is integrated in
 375 the same manner as the CTL, except that $\bar{\theta}$ used in the calculation of wind stress curl (Eq. (2) and
 376 Eq. (3) is replaced by its time-averaged value derived from the CTL. As ocean currents in the
 377 NoDYN experiments do not vary in time, coupling processes mediated by ocean dynamics are
 378 completely eliminated. Second, in the "NoTHERM" experiments, we remove thermodynamical
 379 interbasin coupling by setting terms involving j in Eq. (4) to their corresponding climatological
 380 values (i.e., replace \bar{T}_j with the time-averaged derived from the CTL experiment and drop ξ_j).

Table 1: Parameters used for the conceptual model experiments proposed by GC01.

L_x	Zonal width of domain	2.75×10^7 m
L_y	Meridional width of domain	1×10^7 m
L_{x1}	Zonal width of basin 1	8.25×10^6 m
L_{x2}	Zonal width of basin 2	4.95×10^6 m
H_1	Mean thermocline depth of basin 1	300 m
H_2	Mean thermocline depth of basin 2	300 m
R_1	Deformation radius of basin 1	5.0×10^4 m
R_2	Deformation radius of basin 2	5.0×10^4 m
δ_1	Width of western boundary layer in basin 1	1.0×10^5 m
δ_2	Width of western boundary layer in basin 2	1.0×10^5 m
D	atmospheric scale height	1.0×10^4 m
λ	Damping coefficient of heat flux	$50 \text{ W} \cdot \text{m}^2 \cdot \text{K}^{-1}$
ρ_o	Reference density of atmosphere	$1.25 \text{ kg} \cdot \text{m}^{-3}$
ρ_w	Reference density of seawater	$10^3 \text{ kg} \cdot \text{m}^{-3}$
C_{pa}	Specific heat of atmosphere	$1.0 \times 10^3 \text{ J} \cdot \text{K}^{-1} \cdot \text{kg}^{-1}$
C_{pw}	Specific heat of seawater	$4.0 \times 10^3 \text{ J} \cdot \text{K}^{-1} \cdot \text{kg}^{-1}$
ν_0	Eddy relaxation parameter	$5.0 \times 10^{-7} \text{ s}^{-1}$
ϵ	Horizontal diffusivity	$1.0 \times 10^3 \text{ m}^2 \cdot \text{s}^{-1}$
$F_0 - A$	Heat flux parameter	$37.5 \text{ W} \cdot \text{m}^{-2}$
B	Heat flux parameter	$2.5 \text{ W} \cdot \text{m}^{-2}$
F	Heat flux parameter	$125 \text{ W} \cdot \text{m}^{-2}$
S	Atmospheric stability	$5.0 \times 10^{-3} \text{ K} \cdot \text{m}^{-1}$
f_0	Coriolis parameter	10^{-4} s^{-1}
β	Planetary beta	$2.0 \times 10^{-11} \text{ s}^{-1}$
d_e	Harmonic average of d and D	3.68×10^3 m
σ_1	Amplitude of gaussian noise forcing (basin 1)	9.0
σ_2	Amplitude of gaussian noise forcing (basn 2)	9.0
σ_3	Amplitude of gaussian noise forcing (zonal mean)	5.0

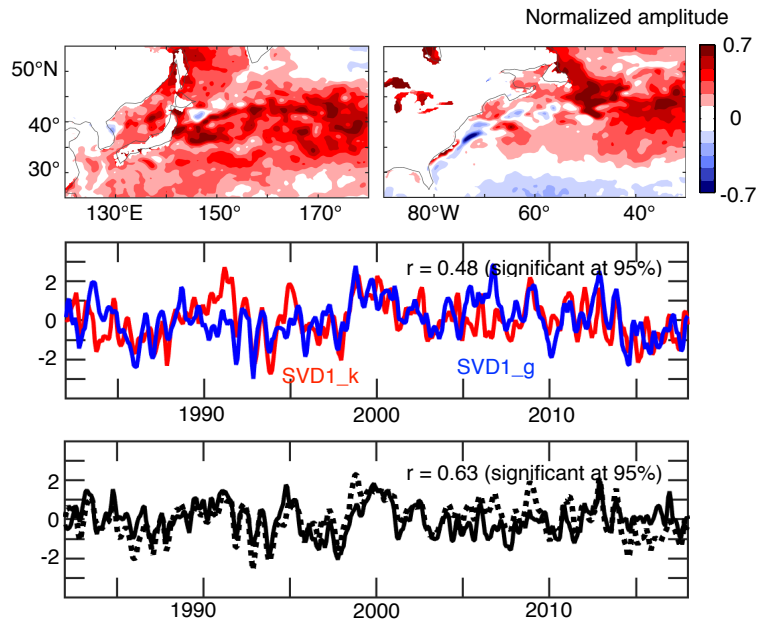
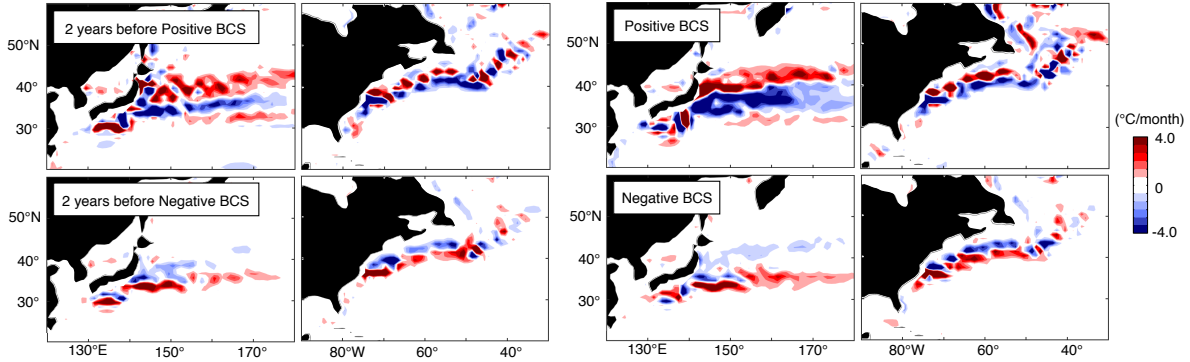


Figure S1: As in Fig. 2, but for observations.

a) Linear dynamical temperature advection anomalies by the geostrophic current, MIROC6subhires



b) Net downward heat flux anomalies in unit of temperature, MIROC6subhires

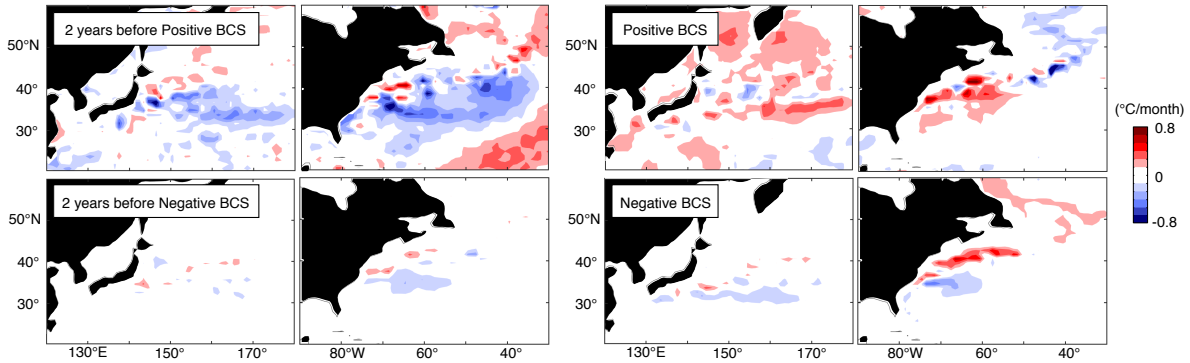


Figure S2: Top, Correlation coefficients between each variable at all latitudes and the Pacific SST anomalies (SST1), $\overline{T_{s1}}$, at 39°N. The left, middle, and right panels show the results from the CTL, NoDYN, and NoTHERM experiments, respectively. Bottom, Time series of the Pacific SST anomalies at 39°N (red), Atlantic SST anomalies at 39°N (blue), and the zonally-symmetric wind stress anomalies at 47.2°N (black). Each time series is linearly detrended and normalized by its own standard deviation. Three experiments are shown in the same manner as in the top panels.