# 70-Year Trends in Ship-Reported Oceanic Precipitation Frequency

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#### Abstract

Ship present-weather reports from 1950 through 2019 are used to assess trends in the reporting of precipitation occurrence over the global oceans. Annual reported precipitation frequency shows statistically significant positive trends of up to  $\sim15\%$  per decade throughout most ocean areas equatorward of 45 degrees. However, latitudes poleward of 45 degrees are dominated by negative trends, some areas of which meet the 95\% confidence threshold. Nine smaller regions were subjectively selected for further investigation, revealing that the observed trends, both positive and negative, are often but not always nearly linear, with the amplitude of interannual fluctuations usually being much larger than that expected from random sampling error alone. The annual time series reveal that four comparatively dry areas are associated with the largest overall positive trends, ranging from 8.3\% to 12.8\% (relative) per decade. Trends were also computed separately for each season, revealing remarkable overall consistency in trends across seasons.











# 70-Year Trends in Ship-Reported Oceanic Precipitation Frequency

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# **5 Key Points:**

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- Significant long-term trends are found in ship reports of precipitation occurrence.
- Trends are mostly positive equatorward of  $45^{\circ}$  and negative at higher latitudes.
- Reporting biases that could explain the trends cannot be ruled out but have not
   been identified.

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#### 10 Abstract

Ship present-weather reports from 1950 through 2019 are used to assess trends in 11 the reporting of precipitation occurrence over the global oceans. Annual reported pre-12 cipitation frequency shows statistically significant positive trends of up to  $\sim 15\%$  per decade 13 throughout most ocean areas equatorward of 45 degrees. However, latitudes poleward 14 of 45 degrees are dominated by negative trends, some areas of which meet the 95% con-15 fidence threshold. Nine smaller regions were subjectively selected for further investiga-16 tion, revealing that the observed trends, both positive and negative, are often but not 17 always nearly linear, with the amplitude of interannual fluctuations usually being much 18 larger than that expected from random sampling error alone. The annual time series re-19 veal that four comparatively dry areas are associated with the largest overall positive 20 trends, ranging from 8.3% to 12.8% (relative) per decade. Trends were also computed 21 separately for each season, revealing remarkable overall consistency in trends across sea-22 sons. 23

#### <sup>24</sup> 1 Introduction

In recent years, calibrated satellite measurements have improved our understand-25 ing of global precipitation distribution and seasonal evolution, including that of ocean 26 precipitation (Skofronick-Jackson et al., 2017). However, assessing long-term trends in 27 oceanic precipitation remains challenging due in part to the comparatively short and het-28 erogeneous satellite record (Nicolas & Bromwich, 2011). This is particularly true prior 29 to the advent of operational passive microwave imagers in 1987 as well as up to the present 30 at higher latitudes, where microwave sensors may miss shallower, lighter, and especially 31 frozen precipitation (Panegrossi et al., 2022). 32

Gu and Adler (2022) have undertaken an analysis of trends in precipitation amount covering the 42-year period from 1979 to 2020. Based on the Global Precipitation Climatology Project (GPCP) precipitation product (Adler et al., 2018), their findings reveal a generally weak but statistically significant long-term trend in global mean precipitation. On regional scales, both positive and negative trends have been observed, but statistical significance could not be established.

Non-satellite-based attempts to estimate climatological oceanic precipitation have
 of necessity relied on subjective and qualitative reports of precipitation occurrence and

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type submitted by sparsely and unevenly distributed commercial and military vessels. 41 Researchers produced climatologies of monthly precipitation amounts by assigning nom-42 inal intensities to each common present-weather code and aggregating ship reports over 43 time (Tucker, 1961; Reed, 1979; Dorman & Bourke, 1979; Legates & Willmott, 1990). 44 Island weather stations believed to be representative of open-ocean conditions have also 45 been utilized to estimate ocean precipitation amounts (Morrissey et al., 1995), but these 46 are primarily found in atolls in the western tropical Pacific, leaving the vast majority of 47 the global oceans without quantitative measurements. 48

Sidestepping the challenge of estimating rainfall amount from categorical ship weather 49 reports, Petty (1995) derived a global climatology of ocean precipitation frequency, also 50 known as fractional time precipitating, and precipitation characteristics from 34 years 51 of synoptic ship present-weather reports spanning the period 1958 to 1991. While the 52 categorization of precipitation type and intensity in these reports is inherently subjec-53 tive, the determination of whether or not it is precipitating is far less so. The study aimed 54 to evaluate satellite-based determinations of simple precipitation occurrence and to elu-55 cidate regional and seasonal variations in precipitation properties that could introduce 56 biases into satellite retrievals of rainfall amount. The high-latitude ocean precipitation 57 frequencies derived by Petty (1995) were in sharp contrast with the lower passive microwave-58 derived estimates of that era (Petty, 1997), but were later largely corroborated by Cloud-59 Sat observations (Ellis et al., 2009). 60

Here we take a first step toward updating and extending the previous analysis by 61 examining trends in precipitation frequency over the 70-year time period from 1950 to 62 2019. This effort is motivated in part by model projections of changes in precipitation 63 amount and distribution in response to climate change (Trenberth, 1999; Chou et al., 64 2012). One must distinguish, however, between precipitation amount and precipitation 65 frequency, as the former is the product of precipitation frequency with the mean non-66 zero precipitation rate, which may itself change in a changing climate (Bichet & Died-67 hiou, 2018). Also, determinations of precipitation frequency are sensitive to the tempo-68 ral window employed—e.g., instantaneous, hourly, daily, etc. (Trenberth & Zhang, 2018). 69 The precipitation frequency examined herein reflects human observations of sky and weather 70 typically lasting less than 10 minutes and is unlikely to differ much from instantaneous 71 determinations except perhaps where extremely intermittent showery precipitation pre-72 vails. 73

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Figure 1. a) Gross annual counts and selected subsets of ship weather reports utilized in the analysis. b) Percentage of reports of significant present-weather ( $ww \geq 30$ ) and of precipitation ( $ww \geq 50$ ) relative to all reports with non-missing sky cover reports N.

#### 74 2 Data

#### 75 **2.1 Source**

We used the latest version of the International Comprehensive Ocean-Atmosphere 76 Data Set (ICOADS) Release 3, Individual Observations (Freeman et al., 2017), which 77 is available through 2019 and continues to be updated. Although this data set includes 78 everything from manned vessels and buoys to autonomous profiling devices and tide gauges, 79 the specific platform types associated with human observations of present-weather in-80 clude "U.S. Navy" (22.1 million reports, 1950–2019), "merchant/foreign ship" (2.9 mil-81 lion), "ocean station vessel – off station" (0.5 million), "ocean station vessel – on sta-82 tion" (0.9 million), "lightship" (1.0 million), and, the largest category, generic "ship" (100.9 83 million). 84

Of the above platform types, only reports from type "ship" are available without interruption throughout the period of interest. Reports identified as "U.S. Navy" are the most numerous type during the first decade of the period but abruptly disappear from the record starting in about 1978, only to reappear in moderate numbers after 2004. Reports from "merchant/foreign ship" comprise about 2.5% of the total and are also un-

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evenly distributed through the record. Because U.S. Navy ships are known to have continued transmitting weather reports during years where they do not appear as such in
the ICOADS record (the first author was a Navy shipboard weather observer from 1978–
1980) and because there was no sudden drop in the total annual report count from all
sources, we surmise that Navy reports were tagged as platform type "ship" during the
missing period. We ultimately elected to utilize the above three platform types in our
analysis, reserving ocean station vessels for possible future use as independent validation at those stations' locations.

Of the 126 million initially qualifying reports, about 1.1 million, or slightly less than 98 1%, are traceable to the tuna fishing fleet of the Inter-American Tropical Tuna Commis-99 sion (IATTC) between 1972–1997 (Smith et al., 2016). These observations, contained 100 in ICOADS deck 667, are mostly found in the eastern tropical Pacific Ocean (Worley et 101 al., 1992) but are usually concentrated in small regions that move about from year to 102 year, presumably following tuna populations. We found that these observations intro-103 duced large temporal and spatial inhomogeneities in both sampling density and reported 104 precipitation frequency, almost always in otherwise data-sparse areas. A previous study 105 by Woodruff (1995) attributed a bias towards weaker winds within the IATTC data to 106 fair weather bias and excluded IATTC data from published enhanced statistics concern-107 ing winds for COADS Release 1a. IATTC data continues to be selectively excluded from 108 COADS Monthly Summary Group products due to these apparent biases (ICOADS, 2016). 109 For similar reasons, as discussed by Freeman et al. (2017), ICOADS observations asso-110 ciated with the Russian Marine Meteorological Data Set (MORMET; deck 732, 7.5 mil-111 lion observations, or about 7% of the total) were excluded from the analysis due to their 112 introduction of large temporal and spatial inhomogeneities. 113

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#### 2.2 Interpretation and quality control

The human-observed present-weather code ww is the primary element of interest in ship synoptic reports, typically taken every 3 or 6 hours. The codes 30 and greater describe "significant" present weather at the location of the station and time of the observation, as distinct from phenomena observed from a distance or during the previous hour. Codes 50 and higher refer to various manifestations of precipitation, with 50–59 being drizzle, 60–69 continuous or intermittent rain, 70–79 frozen precipitation, 80–89 showery precipitation, and 90–99 thunderstorms in progress (Petty, 1995). Any value

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of 50 or higher is thus treated as human observation of precipitation, while all other values are associated with other weather phenomena. Because precipitation takes priority over other possible present-weather elements, a *ww* code value less than 50 rules out the occurrence of precipitation at the time of the observation.

After 1 January 1982, a rule change allowed synoptic reports to omit the ww code 126 if there was no significant weather to report (Dai, 2001). This rule change also introduced 127 the station/weather indicator code ix, which discriminated between manned and auto-128 mated observations (excluded here) and whether or not ww was omitted due to a lack 129 of significant weather or rather due to a lack of data. However, evident inconsistencies 130 in the reporting of the *ix* flag precluded its use as the means to distinguish between miss-131 ing ww and lack of significant present weather. Instead, we interpreted the appearance 132 of non-missing present-weather ww or sky cover N in a report as evidence that a human 133 observer had made a sky and present-weather observation, in which case missing ww likely 134 implied no significant weather. 135

The final dataset contained 103.7 million reports after applying platform type and deck exclusions as well as the test for non-missing ww or N. This total is equivalent to an average of 507 ships reporting every 3 hours over the 70-year record. The annual report counts are depicted in Fig. 1a, including the total (top curve) as well as subsets broken out according to whether N and/or ww were reported. There are clearly large variations in both the total number of reports available and in the proportions of different subsets.

To assess whether these obvious heterogeneities might spill over into computed pre-143 cipitation frequencies, we examined the global percentage of ww reports with values of 144 30 or greater (significant present weather) and 50 or greater (precipitation at the time 145 of the observation). Both fractions are free of large fluctuations over most of the record 146 (Fig. 1b) but exhibit a positive trend of approximately 4.5% (relative) per decade un-147 til about 2007, after which there is a rather sharp fall-off in both fractions. The latter 148 period coincides with a marked low point in the overall availability of reports, and it re-149 mains unclear pending further investigation whether the decline in apparent significant 150 weather frequency reflects a new reporting bias or rather a shift in the geographic dis-151 tribution of available reports. Changes in local and regional reported precipitation fre-152

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quency discussed below should be viewed with caution after 2007, especially if inconsis tent with the previous 58 years.

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#### 2.3 Potential biases and sampling limitations

Potentially more problematic are inconsistencies or biases in operational procedures, 156 especially those that might change over the 70-year record. For example, a fair-weather 157 bias can arise if ships change course to avoid storms, while a foul-weather bias can oc-158 cur if weather reports are only submitted when weather is deemed significant. There may 159 also be differences in reporting practices between merchant and military vessels and be-160 tween crews with varying levels of commitment to World Meteorological Organization 161 reporting standards. It is important to note that shipboard synoptic observations were 162 historically taken to support near-real-time weather analyses of otherwise data-sparse 163 ocean areas and not with long-term climatological applications in mind. 164

For the present purpose, the most significant reporting biases would be those that change over time. However, any such evolution not tied to documented rule changes would be challenging to identify. Therefore, this study relies on less direct evidence of reporting consistency, such as the temporal and/or spatial coherence of computed trends.

Sampling is the single most critical limitation. Figure 2a depicts the total number of included ship reports per 5-degree latitude/longitude gridbox over the entire period. While heavily traveled areas of the north Pacific and Atlantic oceans have frequent reports, reports are scarce over most of the extratropical southern oceans. Figure 2b depicts total counts of precipitation-only reports ( $ww \ge 50$ ), which is the figure most relevant to estimating precipitation frequency. In dry regions where precipitation reports are rare, determining trends can be difficult even if reports are common overall.

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# 2.4 Mean precipitation frequency

Figure 2c depicts the ratio of the precipitation counts in Figure 2b to the report counts in Figure 2a, thus providing a gross (and seasonally biased) depiction of overall precipitation frequency. The magnitudes and spatial patterns are remarkably similar to those derived by Ellis et al. (2009) (their Fig. 3a) using CloudSat observations over a one-year period (August 2006 through July 2007). That such dissimilar data sources and time periods nevertheless yield nearly indistinguishable large-scale distributions of ocean <sup>183</sup> precipitation frequency strengthens the case for the validity of the ship data set for trend<sup>184</sup> analysis.

#### $\mathbf{185}$ **3** Methods

Ship reports were tabulated at 1-degree and monthly resolution to obtain the number of precipitation reports M and total reports N. These initial gridded maps of M and N were then further aggregated over 3-month seasons and coarser spatial resolutions of  $3^{\circ}, 5^{\circ}, 7^{\circ}, 9^{\circ},$  and  $11^{\circ}$  latitude and longitude, as well as a coarsest spatial resolution of  $13^{\circ}$  latitude by  $26^{\circ}$  longitude.

The determination of an unbiased estimate  $\hat{f}$  of the unknown true fraction f of pre-191 cipitation from M and N, along with associated sampling uncertainty, is less trivial than 192 commonly assumed. In particular, for small M, the ratio M/N systematically under-193 estimates the true fraction f for any f > 0. While analytic treatments of this problem 194 exist, we opted to use a Monte Carlo-generated lookup table to obtain an unbiased es-195 timate of not only f but also the sampling uncertainty  $\sigma$  as functions of M and N, given 196 an a priori uniform distribution of f from 0 to 20%. For large  $M, \hat{f} \to M/N$  and  $\sigma \to \sigma$ 197  $\sqrt{M}/N.$ 198

Starting with the coarsest resolution, the estimates  $\hat{f}$  and  $\sigma$  for each 1° gridbox were progressively replaced with results from the next finer resolution if and only if sampling at the new resolution was sufficient to avoid degrading the relative uncertainty  $\sigma/\hat{f}$ . This compositing approach results in final seasonal and annual maps of  $\hat{f}$  and associated  $\sigma$ that are based on coarser-resolution aggregations of reports in data-sparse regions but finer resolution within heavily sampled shipping lanes.

For trend determination, we utilized ordinary least-squares regression (OLS). A twotailed Student's *t*-test with 95% confidence level was used to assess the significance of the trends relative to the null hypothesis of zero trend (Fig. 3b). To account for temporal autocorrelation, the effective independent sample size N' assumed for the *t*-test at each location was reduced to N' = N(1 - r)/(1 + r), where N = 70 and r is the lag-1 autocorrelation (Box et al., 2015).

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# 211 4 Results

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#### 4.1 Trends in Annual Precipitation Fraction

Annual trends expressed in percent per decade relative to the mean reported precipitation frequency over the entire period are depicted in Fig. 3a along with areas for which the trend is different from zero with 95% confidence (p < 0.05) in Fig. 3b. The result is a remarkably coherent pattern of large, statistically significant positive trends throughout most ocean areas equatorward of 45°. Maximum positive trends exceed 10% per decade in many areas and approach 15% for portions of the south central Atlantic.

Within the same latitude zone, limited areas of negative trend are seen only over the northwestern Pacific, in the general vicinity of Australia, and within the dry zones just off the west coast of African, near 15°N and 15°S, as well as near 5°S off the Peruvian coast. However, most of these areas of negative trend do not meet the chosen significance threshold except in the vicinities of Japan and New Zealand.

Latitudes poleward of 45° are dominated by negative trends, some areas of which meet the significance threshold. The latter include much of the North Atlantic and parts of the Barents Sea.

Irrespective of the locally computed *p*-value for the trends, the high degree of spatial coherence of both positive and negative trends speaks against these being the result of statistical flukes due to sampling noise. While spurious trends could potentially result from variable seasonal patterns of ship traffic, it is seems unlikely that this mechanism could give rise to trends of similar sign and magnitude over such extensive contiguous areas.

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#### 4.2 Annual and seasonal trends in selected areas

Nine smaller regions were subjectively selected for further investigation, as indicated by the dashed boxes in Figs. 2 and 3. The specific box locations were influenced in part by the existence of locally higher sample densities, though "Niño 3.4" (5°S–5°N, 120°W–170°W) was chosen for its association with the El Niño-Southern Oscillation (ENSO) (Barnston, 1997).

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# 4.2.1 Time series of annual frequencies

Time series of the annually aggregated data are depicted as solid curves in Fig. 4. For comparison, the random sampling error  $\sigma$  is shown as dashed curves. Plots are annotated with the means, trends, and *p*-values. In addition, both annual and seasonal trends are reported in Table 1, with trends passing the significance test highlighted in bold.

The time series reveal that the observed trends, both positive and negative, are of-244 ten but not always quasi-linear, with the amplitude of interannual fluctuations often be-245 ing considerably larger than that expected from random sampling error alone. Compar-246 atively rain-free areas ( $\bar{f} \leq 0.04$ )—Niño 3.4, the Gulf of Mexico, the Arabian Sea, and 247 the southeast Atlantic off the coast of southern Africa (Fig. 4a, c, h, and i, respectively)— 248 are associated with largest overall positive trends, ranging from 8.3% to 12.8% per decade. 249 In particular, the reported frequency of precipitation over in the southeast Atlantic box 250 averaged around 1.5% between 1950 and 1970 but then doubled to an average of about 251 3% after 2000. 252

A long-term positive trend is apparent in the Niño 3.4 region along the tropical central Pacific (Fig. 4a). Large peaks in precipitation frequency appear to correspond with El Niño events such as in 1982–83, 1997–98, and 2015–16, while deficits in precipitation frequency appear to correspond with La Niña events such as in 1988–89 and 1998-99. The general correspondence of this time series with well-documented ENSO activity lends further confidence in the ship record for examining precipitation variability.

In the western tropical Pacific (Fig. 4e), an overall linear trend is less apparent; rather, 259 the frequency averages around 6.5% in the first decade, is flat or even slightly decreas-260 ing with a mean of around 8% until 1990, and then increasing fairly sharply to an av-261 erage of over 10% during 2009–2014 before falling off again. Similarly, in the Southern 262 Pacific Convergence Zone (SPCZ; Fig. 4f), there is a general downward trend until the 263 mid-1980s followed by a positive trend ending with a large jump in 2014. The latter jump 264 coincides with a much smaller-than-normal sample size for that year, so that feature may 265 not be reliable. 266

Significant negative trends of 3.0% and 2.4% per decade are seen in the north Atlantic (Fig. 4b) and near the Sea of Okhotsk (Fig. 4g), respectively. In the former case,
the trend is relatively flat until about 2005, after which the precipitation frequency drops

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Table 1. Trends in precipitation frequency for the indicated focus areas, expressed as the percent change per decade relative to the mean for the 70-year period. Results are given for annual reports and for the indicated 3-month seasons. Bolded values are significant at the 95% confidence level.

Region	Latitude	Longitude	Annual	MAM	JJA	SON	DJF
Niño 3.4	5S-5N	170W-120W	+9.4	+9.5	+9.3	+6.0	+9.5
North Atlantic	45N-60N	45W–30W	-3.0	-3.2	-2.5	-3.1	-1.8
Gulf of Mexico	15N-30N	95W-80W	+10.2	+7.4	+12.5	+11.1	+9.1
Tropical Atlantic	5S-10N	45W–30W	+9.9	+8.9	+10.1	+9.8	+11.4
West. Trop. Pacific	5S-15N	120E-155E	+5.1	+6.9	+4.5	+2.3	+6.5
S. Pac. Conv. Zone	30S-15S	180W - 150W	+3.6	+5.4	+2.8	+2.2	+2.8
Sea of Okhotsk	40N-55N	145E–160E	-2.4	-2.7	-4.9	-3.5	-2.9
Arabian Sea	5N-20N	55E-70E	+8.3	+3.9	+13.2	+8.1	+0.2
Southeast Atlantic	35S-20S	0 - 15E	+12.8	+14.2	+15.3	+ <b>11.0</b> .	+7.1

by about 2.5% in absolute terms or more than 15% relative to its previous average value
of about 16%. It must be noted that the period of largest dropoff roughly coincides with
the global drop in the reporting of significant present-weather seen in Fig. 1, so we cannot rule out a reporting bias contributing to the dropoff in Fig. 4b. Hints of similar declines are seen in certain of the other time series, including two with otherwise strong
positive trends, such as the Gulf of Mexico (Fig. 4c) and the western tropical Pacific (Fig. 4c).

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# 4.2.2 Trends by season

Within the geographic boxes described above, ship reports were further stratified into 3-month periods—March/April/May (MAM), June/July/August (JJA), September/October/November (SON), and December/January/February (DJF) to permit the determination of trends separately within each season. This also reduces the potential for intraannual sampling biases in the determination of trends. Trends for each period are given in Table 1.

The most striking overall result is that for almost all geographic boxes, not only the sign but also the general magnitude of the trend is similar across all four seasons. This high degree of consistency appears to rule out statistical sampling error as the source of the apparent trends. It also suggests that whatever meteorological or procedural changes may have occurred over the 70 years, they are not significantly influenced by time of year.

#### **5** Conclusions

Our initial analysis of 70 years of shipboard synoptic weather reports reveals significant positive trends in oceanic precipitation occurrence over broad swaths equatorward of 45° latitude, but predominantly negative at higher latitudes. We have not identified any potential sampling or reporting bias that could give rise to the observed geospatial patterns and general consistency across seasons. Unfortunately, there exists no ocean precipitation dataset both extensive and homogeneous enough to validate our findings globally, though local and regional comparisons may be possible.

If real, the positive trends at lower latitudes are not inconsistent with those expected 296 due to global warming and associated mechanisms (Chou et al., 2012), while negative 297 trends at higher latitudes might be related to reduced open-cell convective precipitation 298 and/or precipitation suppression due to increasing anthropogenic aerosol (Rosenfeld et 299 al., 2006). Further analysis is also needed to assess the relationship between these trends 300 and known interannual and interdecadal climate variations (Gu & Adler, 2013). Finally, 301 ongoing work with this dataset includes assessing trends in the fractions of precipitation 302 due to drizzle, snow, and other subclasses as well as undertaking intercomparisons with 303 ocean station vessels, where available, and with the much shorter record of satellite-derived 304 precipitation. 305

#### <sup>306</sup> 6 Open Research

ICOADS-3 data are available from Research Data Archive, Computational and Information Systems Laboratory, National Center for Atmospheric Research, University Corporation for Atmospheric Research et al. (2016). Complete Jupyter/Python notebooks used to obtain numerical and graphical results herein are posted at github.com/ gpetty/GRL-2023.

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Figure 2. a) Total counts of included ship weather reports ("U.S. Navy", "merchant/foreign ship", and "ship" platform types) per  $5^{\circ} \times 5^{\circ}$  degree box over the 70-year period of interest. b) Counts of reports indicating present precipitation only. c) The ratio of precipitation reports to total reports. Nine dashed boxes depict areas selected for additional analysis.



Figure 3. a) Mean trend over 70 years in precipitation frequency computed from ship reports aggregated by year. Trends are expressed as percent changes (relative to the mean precipitation frequency) per decade. Gray regions denote areas lacking the minimum number of ship observations for at least 5 years in the 70-year record. b) Areas with trends different from zero with 95% confidence. Nine dashed boxes depict areas selected for additional analysis.



**Figure 4.** Time series of precipitation frequency computed from yearly ship reports within the indicated latitude/longitude boxes, which correspond to those depicted in Figs. 2 and 3. Dashed lines indicate the random sampling uncertainty.

Figure 1.



Figure 2.



b) Reports of Precipitation (1950-2019)





Figure 3.



a) Trend in Ship-Reported Ocean Precipitation Frequency

b) Trend Different from Zero With >95% Confidence



Figure 4.

