

1 **Supporting Information for “Increases in Future AR**
2 **Count and Size: Overview of the ARTMIP Tier 2**
3 **CMIP5/6 Experiment”**

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27 **Introduction**

28 This supplemental information provides additional useful details on ARDTs, their
29 treatment of thresholds, and our grouping of ARDTs into categories. The supplemental
30 figures expand on figures in the main text to show all ARDT-simulation combinations.

31 **Text S1.**

32 **Treatment of Thresholds**

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33 We document here choices/specializations (if any) that ARDT contributors made in
34 running their ARDTs on the Tier 2 CMIP5/6 simulations

- 35 • `ARCONNECT_v2`: only uses absolute threshold; no Tier 2-specific decisions needed
- 36 • `Guan_Waliser_v2`: uses 85th percentile from the historical simulation
- 37 • `IDL_rel_future`: uses 85th percentile calculated from the future simulation
- 38 • `IDL_rel_hist`: uses 85th percentile calculated from the historical simulation
- 39 • `Lora_v2`: uses a time-and-latitude dependent IVT threshold that asymptotes to 225
40 kg/m/s at the poles; the time/latitude dependence of the threshold is a function of the
41 30-day running mean and a zonal average of IWV, so no Tier 2-specific decisions are
42 needed

- 43 • `Mundhenk_v3`: calculates the mean and seasonal cycle of IVT based on the historical
44 simulation and removes this to determine the IVT anomaly relative to the historical period

- 45 • `PNNL_v1`: only uses absolute threshold; no Tier 2-specific decisions needed
- 46 • `TECA_BARD_v1.0.1`: uses threshold relative to spatial map of IVT at a given time;
47 no Tier 2-specific decisions needed

48 The `Mundhenk_v3` algorithm differs from prior published versions (i.e., `Mundhenk_v1`,
49 `Mundhenk_v2`) in its more reliable detection of AR objects that cross the boundary of the
50 dataset's spatial domain.

51 The `Tempest` ARDT uses an absolute threshold for the laplacian of IVT. The Tier
52 1 version also utilized an absolute threshold of 250 kg/m/s of IVT, but it was later
53 determined that this threshold had no effect on the ARDT results because regions that
54 satisfied the Laplacian threshold also satisfied the IVT threshold. The minimum latitude
55 for ARs was raised to 20°, from 15°, to filter easterly waves. The stencil radius and

56 magnitude used for the Laplacian depends on the model grid, and this is held constant
57 for the historical and future simulations.

58 Discussions with the **Tempest** contributing scientists indicate that the algorithm may
59 benefit from further tuning of their method when applied to moderately low resolution
60 data, and efforts are underway to provide a second version of their contribution to Tier
61 2. Such discoveries and improvements are a benefit of intercomparison projects.

62 **Text S2.**

63 **Classification of ARDTs**

64 Building on Rutz et al. (2019), we classify the Tier 2 CMIP5/6 ARDTs into three
65 groups, based on their treatment of thresholds: *absolute*, *fixed relative*, and *relative*. These
66 classifications are indicated as *abs.*, *fix. rel.*, and *rel.* in Table S1. A key motivation for
67 this categorization is aggregating ARDTs by their sensitivity to thermodynamic changes
68 in IVT, with the assumption that ARDTs employing absolute thresholds to moisture fields
69 will be the most sensitive, and ARDTs employing time-dependent thresholds will be least
70 sensitive.

71 **Absolute ARDTs:** We define *absolute ARDTs* as utilizing any fixed thresholds (e.g.,
72 in IVT) for discriminating ARs from the background. **ARCONNECT_v2** and **PNNL_v1** unam-
73 biguously fit in this category. **Lora_v2** uses an IVT threshold that varies with latitude and
74 time, and the threshold asymptotes to 250 kg/m/s at mid-to-high latitudes (the thresh-
75 old increases toward infinity approaching the tropics). This design effectively imposes an
76 absolute threshold of at least 250 kg/m/s. Because of this, we classify **Lora_v2** as an
77 *absolute ARDT*, while recognizing that this is not a perfect categorization.

78 **Fixed relative ARDTs:** We define *fixed relative ARDTs* as those that employ relative
79 thresholds that do not vary with time. For example, `Guan_Waliser_v2` calculates the 85th
80 percentile of IVT from the historical simulations and discriminates ARs from the back-
81 ground where IVT is greater than the local, historical 85th percentile; hence the threshold
82 used in the `Guan_Waliser_v2` algorithm does not change in time. The `IDL_rel_hist`
83 and `IDL_rel_future` ARDTs use a similar approach and are therefore also categorized
84 as *fixed relative ARDTs*. `Mundhenk_v3` calculates IVT anomalies relative to the historical
85 period and identifies ARs that are above the 94th percentile of the historical simulation,
86 so it also fits unambiguously in the *fixed relative* category.

87 **Relative ARDTs:** We define *relative ARDTs* as those that employ relative thresholds
88 that vary with time. `TECA_BARD_v1.01` unambiguously fits into this category, since ARs
89 are identified where IVT is above a fixed percentile of IVT, where the percentile is calcu-
90 lated in space (in contrast to time, e.g., for `Guan_Waliser_v2`). `Tempest` uses an absolute
91 threshold applied to the Laplacian of the IVT field, which might warrant its classification
92 as an absolute ARDT. However, the use of the Laplacian removes the mean of the IVT
93 field; therefore `Tempest` identifies areas of IVT that are high relative to nearby areas of
94 IVT at the same timestep. We therefore classify `Tempest` as a *relative ARDT*.

95 **Text S3.**

96 **Details on Missing Data** All ARDTs detect ARs for the 1951-2099 period for the
97 combined historical and future simulations for each CMIP5/6 model. We analyze output
98 from the entire 1951-2099 timeperiod. There are some exceptions to this: output from the
99 CMIP6 IPSL-CM6A-LR SSP5-8.5 simulation are only available through 2049, there are
100 data corruption issues for the year 2006 in the CMIP5 CSIRO-Mk3-6-0 simulation, and

101 there are data corruption issues for the year 2095-2099 for the TECA_BARD_v1.01 output
102 applied to the CMIP5 IPSL-CM5B-LR simulation. Years with data corruption issues
103 are marked as missing, and trends and climatologies are only calculated considering non-
104 missing data. The Guan_Waliser_v2 algorithm did not supply ARDT catalogues for the
105 NorESM1-M and BCC-CSM2-MR simulations due to technical issues at the time.

106 **Text S4.**

107 **Comment on the Tier 2 Reanalysis Experiment**

108 The tiered structure of the ARTMIP experiments requires that all participants con-
109 tribute to the Tier 1 experiment; by design, all ARDTs participating in the Tier 2
110 CMIP5/6 experiment also have been run on MERRA-2 as part of the Tier 1 experiment.
111 A separate ARTMIP Tier 2 experiment is currently underway, comparing ARDT results
112 applied to different reanalyses. The set of ARDTs participating in the Tier 2 Reanalysis
113 experiment is not identical to the set participating in this Tier 2 CMIP5/6 experiment,
114 so use of multiple reanalyses is not possible for ARDTs. For the sake of uniformity in the
115 experimental approach, we use only the MERRA-2 reanalysis. Preliminary analysis of
116 the Tier 2 Reanalysis experiment (not shown) indicates that the uncertainties associated
117 with choice of reanalysis are small compared to the uncertainties discussed in this paper,
118 and it is therefore unlikely that use of a different reanalysis would change the qualitative
119 conclusions of this paper.

Table S1. (left) ARDT algorithms, and associated metadata, that contributed to the Tier 2 CMIP5/6 experiment. ARDT classifications ('Class.')

(right) Details of CMIP5/6 models used in the Tier 2 experiment.

ARDTs				Models			
Algorithm ID	Contrib.	Class.	Region	MIP Era	Model Name	Inst.	~Res. [km]
ARCONNECT_v2	Shearer	abs.	Global	CMIP5	CCSM4	NCAR	120
GuanWaliser_v2	Guan	fix. rel.	Global	CMIP5	CSIRO-Mk3-6	CSIRO	207
IDL_rel_future	Ramos	fix. rel.	W. Europe, S. Africa	CMIP5	CanESM2	CCCMA	310
IDL_rel_hist	Ramos	fix. rel.	W. Europe, S. Africa	CMIP5	IPSL-CM5A-LR	IPSL	296
Lora_v2	Lora	abs.	Global	CMIP5	IPSL-CM5B-LR	IPSL	296
Mundhenk_v3	Nardi	fix. rel.	Global	CMIP5	NorESM1-M	NCC	242
PNNL_v1	Sarangi	abs.	W. U.S.	CMIP6	BCC-CSM2-MR	BCC	124
Tempest	McClenny	rel.	Global	CMIP6	IPSL-CM6A-LR	IPSL	198
TECA_BARD_v1.01	O'Brien	rel.	Global	CMIP6	MRI-ESM2-0	MRI	124



Figure S1. Maps of AR frequency (AR days per year) from 1981-2010. Columns correspond to ARs detected by specific ARDTs, and rows correspond to input datasets (MERRA-2 for the first row and CMIP5/6 for other rows). The rightmost column shows the multi-ARDT mean frequency for each model. The bottom row shows the multi-model mean for each ARDT (excluding MERRA-2 from the mean). The bottom right panel shows the multi-ARDT mean frequency (excluding MERRA-2 from the mean).



Figure S2. Maps of AR frequency (AR days per year) from 2070-2099. Columns correspond to ARs detected by specific ARDTs, and rows correspond to CMIP5/6 models. The rightmost column shows the multi-ARDT mean frequency for each model. The bottom row shows the multi-model mean for each ARDT. The bottom right panel shows the frequency from the last 30 years of available simulation data (2020-2049).

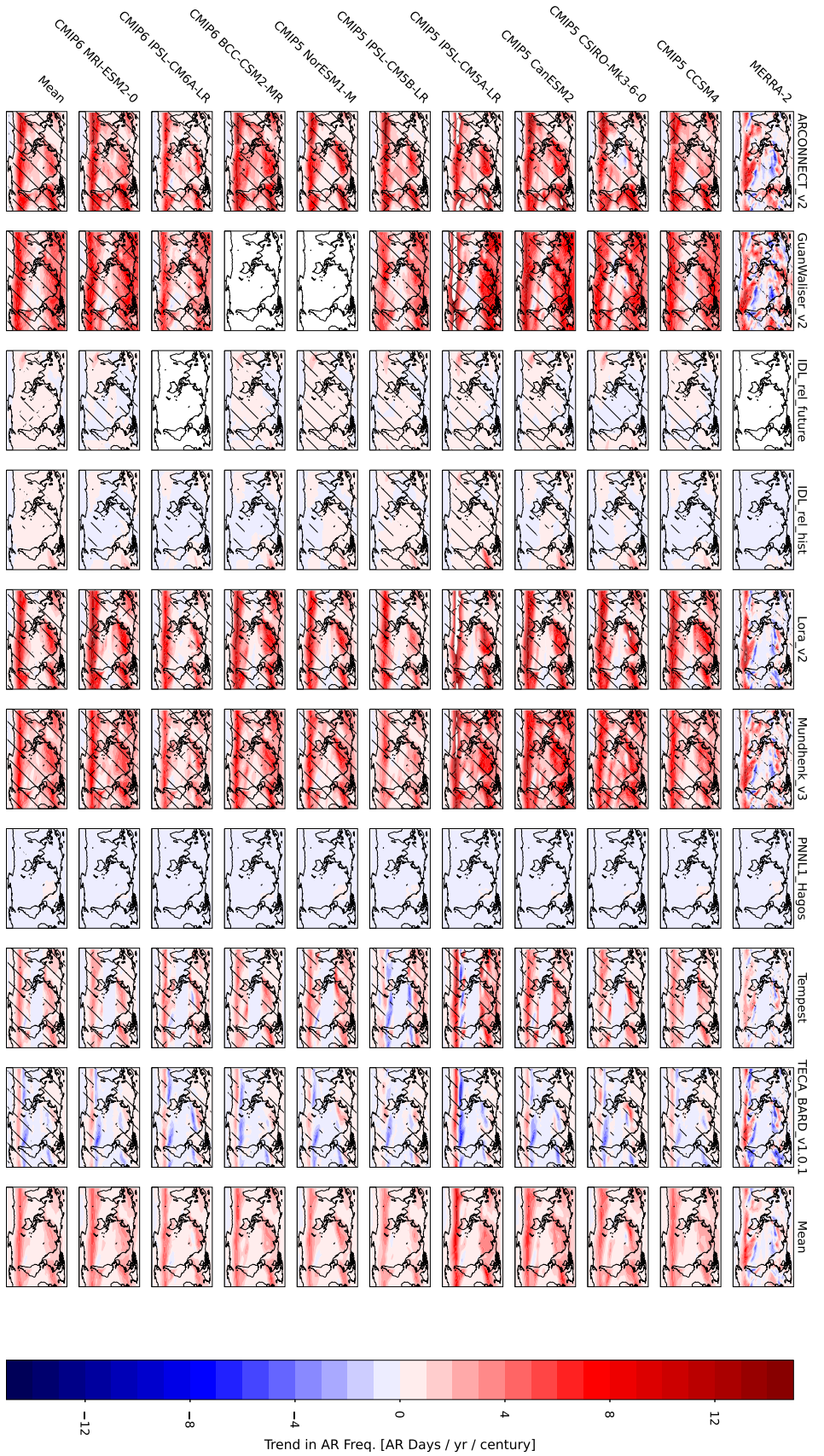


Figure S3. Maps of trends in AR frequency (AR days per year per century) from 1981-2099 (from 1981-2017 for MERRA2).

Columns correspond to ARs detected by specific ARDTs, and rows correspond to input datasets (MERRA-2 for the first row and CMIP5/6 for other rows). The rightmost column shows the multi-ARDT mean trend for each model. The bottom row shows the multi-model mean for each ARDT (excluding MERRA-2 from the mean). The bottom right panel shows the multi-model, multi-ARDT mean frequency (excluding MERRA-2 from the mean). Trends for CMIP6 IPSL-CM6A-LR are calculated through 2049.

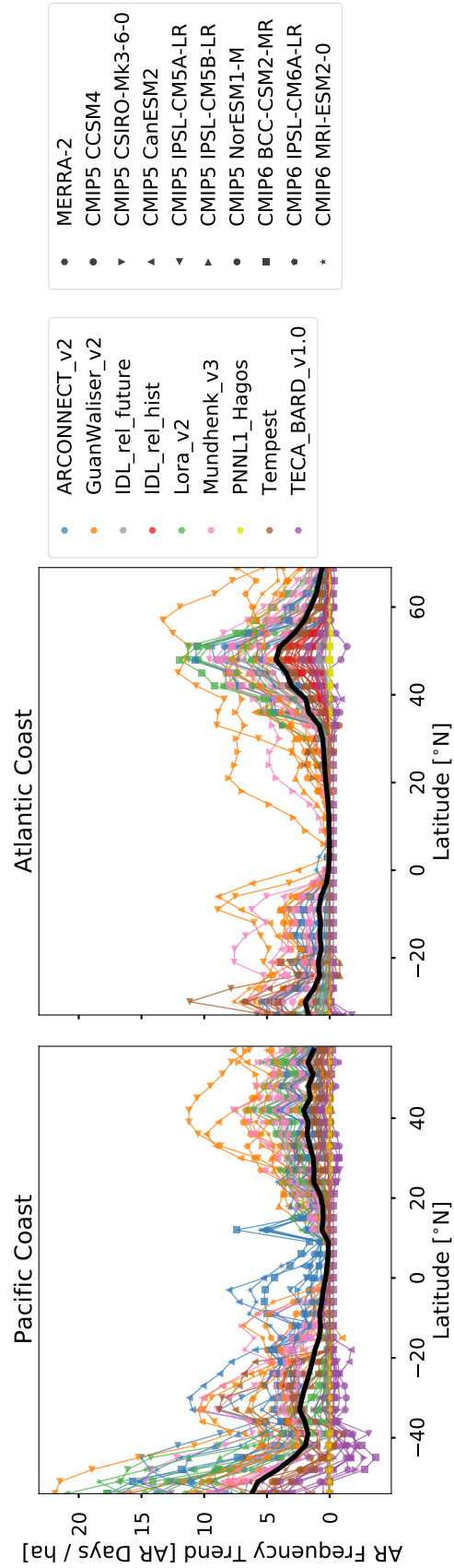


Figure S4. Trends in AR frequency (AR days per year per century) from 1981-2099 for western coastlines along the (a) Pacific and (b) Atlantic oceans. Colors correspond to ARDTs and markers correspond to simulations.

