1	Significant Modulation of Variability and Projected Change in California
2	Winter Precipitation by Extratropical Cyclone Activity
3	
4	Edmund K.M. Chang ¹ , Cheng Zheng, Patrick Lanigan, Albert M.W. Yau
5	School of Marine and Atmospheric Sciences
6	Stony Brook University, Stony Brook, NY
7	
8	J. David Neelin
9	Department of Atmospheric and Oceanic Sciences
10	University of California at Los Angeles, Los Angeles, CA
11	
12	April 2015
13	Geophys. Res. Lett., Accepted June 2015
14	
15	
16	Abstract
17	Extratropical cyclones give rise to much of the precipitation over California. Observed
18	California winter precipitation is highly correlated to a metric of extratropical cyclone activity
19	over the Eastern Pacific. The lack of precipitation over the recent winters is coincident with
20	consecutive winters of much below average cyclone activity. Analysis of variability in cyclone
21	activity and California precipitation simulated by models participating in Coupled Model
22	Intercomparison Project phase 5 indicates that most models can simulate the relationship
23	between cyclone activity and precipitation well. Examination of projected change suggests 1) no
24	evidence of a long-term downward trend in California-region cyclone activity within the
25	examined scenarios; and 2) that the inter-model spread in California precipitation projection can
26	be largely explained by the spread in the projection of extratropical cyclone activity. This
27	highlights the need to further understand physical mechanisms for the variation in projection of

28 cyclone activity in this region.

¹Corresponding author: Dr. Edmund K.M. Chang, School of Marine and Atmospheric Sciences, Stony Brook University, Stony Brook, NY 11794-5000. Email: kar.chang@stonybrook.edu

31 **1. Introduction**

California receives most of its precipitation during winter, and much of that is brought by the passage of extratropical cyclones and their fronts. During the three consecutive winters of 2011/12 through 2013/14, precipitation over California was all much below average, and most of California experienced either extreme or exceptional drought conditions during much of 2014 [Seager et al., 2014a]. Due to the drought, the statewide water storage in November 2014 was only about 56% of average for that time of the year [Seager et al., 2014a].

Lack of precipitation is not the only cause of the drought. Several studies have suggested that higher than average temperatures during these years also contributed to the dryness [e.g. AghaKouchak et al., 2014; Griffin and Anchukaitis, 2014]. Nevertheless, it is still important to understand the physical drivers behind California precipitation variability, as well as better understand future projections derived from climate model simulations.

In projections of precipitation change in the Coupled Model Intercomparison Project 43 (CMIP) Phase 3 and Phase 5 ensembles of model simulations, precipitation reductions in the 44 subtropics and increases at mid- to high-latitudes tend to be relatively well agreed upon at large 45 scales [Meehl et al., 2007; Collins et al., 2013; Maloney et al., 2014; Seager et al., 2014b]. 46 47 California lies close to the boundary between these opposing tendencies [Cayan et al., 2008; Pierce et al., 2012], so precipitation change projections require careful scrutiny of uncertainties 48 that can be aided by an understanding of dynamical factors influencing the region. Neelin et al. 49 [2013] note that precipitation change in Central and Southern California constitutes one of the 50 51 main features of North American end-of-century projections that differs in CMIP5 simulations relative to CMIP3, associated with a change in the storm track precipitation coming onto the 52 coast. This feature is associated with simulated increases in 200 hPa winds in a location that 53 54 tends to extend the region of strong subtropical jet toward the east, which was postulated to tend 55 to extend the storm track toward the California coast. Chang et al. [2012] and Chang [2013] find a tendency of weakening in measures of storm track or extratropical cyclone activity in 56 57 hemispheric and aspects of the large-scale North American response, potentially associated with 58 large-scale changes in baroclinicity. Potential changes affecting the California region are closely 59 associated with the regional dynamics where storms approach the coast. Here, we address the 60 connection between measures of extratropical cyclone activity, in particular bandpass-filtered sea level pressure variance, and the changes in precipitation over California, both in interannual 61 variations and in global warming projections. 62

63

64 **2. Data and Methods**

65 To obtain the climatological extratropical cyclone activity and precipitation, we have used two analysis data sets to represent observations. Extratropical cyclone activity is calculated 66 based on the National Centers for Environmental Prediction-National Center for Atmospheric 67 Research reanalysis (NNR) data [Kalnay et al. 1996] with a horizontal resolution of 2.5°. 68 Precipitation in California is derived from the Global Precipitation Climatology Project (GPCP, 69 version 2.2) monthly data [Adler et al. 2003], which is a gridded analysis merging satellite 70 observations with gauge measurements. The time periods of both NNR and GPCP data examined 71 are from December 1979 to February 2015, 36 winter seasons (December to February, DJF) in 72 73 all.

In total 30 CMIP5 models from 18 modeling centers (Table S1) have been examined in
 this study. These represent all models that have provided 6-hourly data for both the historical and

76 future periods that are currently available from the CMIP5 archives (note that 6-hourly data from 77 FGOALS-s2, examined in some previous studies, have been withdrawn). Extratropical cyclone activity is calculated from 6-hourly data and model precipitation is derived from monthly mean 78 79 data. Similar to Neelin et al. [2013], we selected 40 winter seasons in both historical runs and Representative Concentration Pathway 8.5 (RCP8.5) simulations to represent the climatology in 80 the 20th Century (1961-2000) and the end of 21st century (2059-2098). The high emission 81 82 RCP8.5 scenario is analyzed to provide a strong signal. Note that previous studies have 83 suggested that many climate change signals, including projected changes in extratropical cyclone activity, scales roughly linearly with the hemispheric temperature change [e.g., Neelin et al., 84 85 2006; Chang, 2013; Tebaldi and Arblaster, 2014]. The winter season extratropical cyclone activity and precipitation changes are defined as the difference between the climatology of 86 historical period and RCP8.5 simulations. All the fields in CMIP5 models are interpolated onto 87 the same $2.5^{\circ} \times 2.5^{\circ}$ grid before analyses are performed. Models in each ensemble are treated 88 independently, though some of them may have the same structural cores so they may have 89 similar biases [e.g. Knutti et al 2010, 2013]. Historical simulations by FGOALS-g2 fail to 90 capture the relationship between extratropical cyclone activity and precipitation (see discussions 91 below), hence FGOALS-g2 is not included in multi-model ensemble (MME) mean, and all 92 model results are based on 29 CMIP5 models from 17 modeling centers. 93

Most CMIP5 models only provide 6-hourly data from a single ensemble member. Several 94 models provide data from multiple members, and historical and future experiments from a single 95 ensemble member (usually r1i1p1) are used in the analysis. Since GISS-E2-H and GISS-E2-R 96 only provide 6-hourly historical and future data from different ensemble members, we have used 97 data from different ensemble members (with the same physics) to represent the historical period 98 and RCP8.5 simulations for the two GISS models. CSIRO-Mk3-6-0 has provided the largest 99 number of ensemble members (10), and these 10 members have been analyzed to explore 100 internal climate variability. Results from these 10 members (see discussions below) also confirm 101 that it is appropriate to use different ensemble members to represent historical period and 102 RCP8.5 simulations in our study. 103

Precipitation in California is defined as the winter season precipitation in the shaded area in Fig.1e (same as the California precipitation box in [Neelin et al. 2013]), which covers about the southern two-thirds of California, where the sign of precipitation projection is uncertain. Extratropical cyclone (or storm track) activity is defined based on temporal variance statistics, band-pass filtered using a 24-hour difference filter [Wallace et al. 1988]:

109

$$pp = \overline{\left\{p(t+24hr) - p(t)\right\}^2} \qquad (1)$$

In equation (1), *p* is sea level pressure (SLP), and *pp* is the 24-hour difference filtered variance of sea level pressure. The overbar corresponds to time averaging for the winter season (DJF). Many previous studies [e.g. Lau, 1978; Wallace et al., 1988; Chang et al., 2002] have suggested that the peaks in this measure lie over geographical locations where cyclone tracks preferentially cross (see also Fig. 1).

115

116 **3. Results**

117 **3.1 Historical period**

The climatological distribution of winter extratropical cyclone activity as indicated by *pp* across Eastern North Pacific and North America is shown in Fig. 1a. Maximum activity extends northeastward from Central North Pacific into the Gulf of Alaska and western Canada, and then southeastward across Canada towards the Great Lakes, qualitatively similar to major cyclone tracks that traverse the region [e.g. Hoskins and Hodges 2002]. Our own analyses (not shown) have also shown that this distribution is very similar to the climatological spatial distribution of extratropical cyclone density. Note that the maxima of *pp* are generally located poleward of the mid-latitude maxima in precipitation (not shown here, but see, for example, Fig. 1 of Neelin et al. [2013]). This is because cyclone related precipitation extends far to the equatorward side of cyclone centers along cold and warm fronts [e.g. Chang and Song 2006; Bengtsson et al. 2009].

The multi-model mean *pp* climatology based on historical experiments from CMIP5 models is shown in Fig. 1b. Overall, CMIP5 models are quite successful in simulating the distribution except that the spatial pattern is smoother due to averaging over a large ensemble.

The time series of winter precipitation over the California box (Fig. 1e) based on GPCP data is shown by the blue line in Fig. 2b. There is clearly large year-to-year variability – seasonal precipitation varies from about 1.2 mm day⁻¹ in 1990/91 to about 6.4 mm day⁻¹ in 1997/98. There is a weak decreasing trend of -0.025 mm day⁻¹ year⁻¹ between 1979/80 and 2014/15, but this trend is not significant at the 90% level.

To quantify the relationship between California precipitation and extratropical cyclone 136 activity, point-by-point correlation is computed between the California precipitation time series 137 138 shown in Fig. 2b with the time series of pp at each grid box. The resulting correlation map is shown in Fig. 1c. It is clear that California precipitation is strongly correlated with extratropical 139 cyclone activity over Eastern Pacific just to the west of California, with positive correlation 140 reaching over 0.8. This positive correlation is reasonable since most of the precipitation over 141 California is due to passages of extratropical cyclones or their fronts. There is also a weaker but 142 still statistically significant negative correlation spreading across the south coast of Alaska and 143 144 western Canada, suggesting that the increased cyclone activity over Eastern Pacific that brings more precipitation to California is frequently associated with a south-eastward shift of the East 145 Pacific storm track. Part of this variability is related to the El Niño-Southern Oscillation (ENSO), 146 as previous studies [e.g. Chang et al., 2002] have shown that during El Niño, the Pacific jet is 147 shifted slightly southward and extends eastward towards North America, steering more 148 extratropical cyclones towards California [Held et al., 1989; Straus and Shukla, 1997; Dettinger 149 et al., 1998; Cayan et al., 1999]. However, note that the correlation between winter California 150 precipitation and the Nino-3.4 index is only 0.38 between 1979/80 and 2013/14, indicating that 151 ENSO can only explain part of the relationship between California precipitation and extratropical 152 cyclone activity. 153

Similar correlation maps have been computed for each of the CMIP5 historical simulations using their own precipitation and extratropical cyclone activity time series, and the average of these correlation maps is shown in Fig. 1d. Overall, CMIP5 models are able to reproduce the main structure of the observed correlation map, with large positive correlations just to the west of California, and weak negative correlations to the north.

Given the strong positive correlation between California precipitation and extratropical cyclone activity just offshore, in both observations and model simulations, an East Pacific box (Fig. 1) is defined covering the region of highest correlations to quantify year-to-year variations of extratropical cyclone activity. Note that results discussed below are not sensitive to the exact definition of the box – we have shifted the box in all directions by 5° as well as changed the size of the box, and very similar results are obtained.

165 The time series of extratropical cyclone activity averaged over the East Pacific box, based 166 on NNR, is shown by the red line in Fig. 2b. It is clear that this time series is highly correlated 167 with the California precipitation time series. The values of California precipitation and

extratropical cyclone activity for each winter are plotted against each other in Fig. 2a. The 168 169 correlation between the two quantities is about 0.8. Fig. 2b suggests that similar to California precipitation, extratropical cyclone activity also exhibits a small downward trend during this 170 171 period, but again this trend is not statistically significant. Regardless, the three consecutive winters of low precipitation in the early 2010s (2011/12 through 2013/14) are coincident with 172 three years of low extratropical cyclone activity. Note that the ratio between the two trends 173 shown in Fig. 2b (0.18 mm day⁻¹ hPa⁻²) is consistent with the slope of the regression line 174 between the two quantities based on year-to-year variability (0.17 mm day⁻¹ hPa⁻²; see Fig. 2a). 175

176 Scatterplots similar to Fig. 2a have been derived from historical simulations of each of 177 the CMIP5 models, and they are shown in supporting Fig. S1. Twenty nine of the thirty models 178 examined display significant correlation between California precipitation and extratropical 179 cyclone activity averaged over the East Pacific box. The only exception is FGOALS-g2 (bottom 180 right panel in Fig. S1), which differs so clearly from observed behavior that it is excluded from 181 the analyses below.

Considerable model-to-model variability occurs in these historical relationships (Fig. S1). 182 The correlation varies between a high of 0.93 for CCSM4 to a low of 0.44 for GFDL-ESM2G. 183 The multi-model mean correlation is 0.71. The slope of the regression line also varies from 0.07 184 mm day⁻¹ hPa⁻² for GFDL-ESM2G to 0.24 mm day⁻¹ hPa⁻² for MIROC5, with the multi-model 185 mean slope being 0.16 mm day⁻¹ hPa⁻², consistent to the observed slope (0.17 mm day⁻¹ hPa⁻²) 186 discussed above. To see how much of the variability might be due to internal climate variability 187 rather than to model uncertainties [e.g. Hawkins and Sutton, 2011; Deser et al. 2012], similar 188 analyses have been performed on 10 ensemble members from CSIRO-Mk3.6.0 which use the 189 190 same model forced by the same climate forcings. Scatterplots from these 10 simulations are shown in supporting Fig. S2. Among these 10 simulations, the correlation between California 191 precipitation and East Pacific extratropical cyclone activity varies from 0.72 and 0.85, while the 192 slope of the regression line varies from 0.16 to 0.24 mm day⁻¹ hPa⁻². Comparing results shown in 193 Figs. S1 and S2, while climate variability can give rise to significant variability in the 194 relationship between California precipitation and extratropical cyclone activity, it is unlikely that 195 196 internal climate variability can explain all the model-to-model differences. Overall, most models successfully simulate a tight relationship between California precipitation and East Pacific 197 extratropical cyclone activity. 198

199

200 **3.2 Future projections**

CMIP5 multi-model ensemble mean projected change in precipitation is shown in Fig. 3a, and model agreement is shown in Fig. 3b. Consistent with Neelin et al. [2013], increased precipitation is projected over much of California, including much of the box shown in Fig. 1e. Overall, a majority of the models project precipitation increase over most of the region. Nevertheless, there are significant model-to-model differences in the magnitude of the projected change, with the California box averaged precipitation projected to change by nearly -1 mm day⁻¹ to a bit over +2 mm day⁻¹ (see Fig. 4a).

Projected change in extratropical cyclone activity is shown in Fig. 3c. On average, CMIP5 models project a small but statistically significant increase in extratropical cyclone activity over Eastern Pacific, and a majority of the models project increase over much of the East Pacific box (Fig. 3d), especially over the northwestern part of the box. Averaged over the box, the projections range from a bit less than -4 hPa² to more than +10 hPa². The models also project significant decrease to the north of this region, consistent with the results of Chang et al. [2012] that cyclone tracks over the East Pacific just off the North America coast are projected to shiftsouthward by CMIP5 models.

In Fig. 4a, each model's projected precipitation change over the California box is plotted 216 217 against its projected change in extratropical cyclone activity averaged over the East Pacific box. There is clearly high correlation between these two projected changes, with the inter-model 218 correlation being 0.85. The slope of the regression between these two quantities (0.18 mm day⁻¹ 219 hPa⁻²) is also close to the ensemble mean slope of the year-to-year variability discussed above 220 (0.16 mm day⁻¹ hPa⁻²). Among the 29 models, the two MRI models project the largest increase in 221 both precipitation and extratropical cyclone activity and are apparent outliers. Nevertheless, even 222 223 if we remove results from these two models, the correlation between the other 27 models is still quite high (0.73), with a regressed slope of about 0.15 mm day⁻¹ hPa⁻², again close to the multi-224 model mean regressed slope based on the historical period. 225

Given the strong physical and statistical relationships between California precipitation 226 227 and East Pacific extratropical cyclone activity, we hypothesize that model projected change in California precipitation can be "predicted" by model projected change in extratropical cyclone 228 229 activity. To predict precipitation projection based on projected change in extratropical cyclone activity, for each model, its projected change in extratropical cyclone activity is multiplied by the 230 regression coefficient between California precipitation and extratropical cyclone activity found 231 in its historical simulation. For example, MRI-CGCM3 projects an increase in extratropical 232 cyclone activity of about 10 hPa² over the East Pacific box (Fig. 4a). Based on its historical 233 simulation, the regression slope between California precipitation and East Pacific extratropical 234 cyclone activity is 0.206 mm day⁻¹ hPa⁻² (see Fig. S1). These two quantities are then multiplied 235 together to represent the model's predicted change in California precipitation based on its 236 projected change in extratropical cyclone activity, and the result comes out to be about 2.1 mm 237 day⁻¹. This "predicted" quantity is then plotted against the model's actual projected change in 238 239 California precipitation in Fig. 4b. This procedure is repeated for all models, and the resulting plot is shown in Fig. 4b. 240

Fig. 4b shows that there is a tight relationship between each model's precipitation projection and the prediction based on its extratropical cyclone projection. The correlation between these two quantities is about 0.79, and the slope of the relationship (0.97) is close to 1. Even if we remove the results of the two MRI models, the correlation remains quite high (0.61) and the slope, while smaller, is not statistically different from the slope of the regression that includes the two models.

To assess how much of the model-to-model differences shown in Fig. 4 could be due to 247 climate variability rather than model (or response) uncertainties, projections of changes in 248 precipitation and extratropical cyclone activity based on the 10 ensemble members of CSIRO-249 Mk3.6.0 have been examined, and analyses similar to those presented in Fig. 4 have been 250 conducted on the 10 ensemble members (supporting Fig. S3). Results from these 10 members 251 from the same model under the same forcing show substantial spread, covering about 40% of the 252 spread shown in Fig. 4. Member-to-member differences in projected changes in California 253 precipitation are also well correlated with (and well predicted by) member-to-member 254 differences in projected changes in East Pacific extratropical cyclone activity. Examining the 255 anomalies of each member relative to the ensemble mean, no significant correlation is found 256 between each member's precipitation (or extratropical cyclone activity) anomaly during the 257 historical period with its anomaly in the future, i.e., as expected, internal variability has little 258 memory across the intervening half-century. Comparing historical simulation and future 259

prediction from different ensemble members, these combinations also lie along the same regression lines as those from the same ensemble member (see Fig. S3). This justifies using results from different ensemble members for the GISS models, for which historical and future results from the same ensemble member are not available.

264

265 4. Summary and Discussions

California winter precipitation is shown to be strongly modulated by variability and 266 change in extratropical cyclone activity over East Pacific just off the California coast in 267 reanalysis data and CMIP5 simulations. Analyses of interannual variability in California 268 precipitation and East Pacific extratropical cyclone activity reveal that winter-to-winter 269 variations in precipitation are strongly correlated with variations in extratropical cyclone activity. 270 This is not surprising since much of California's precipitation comes from passage of 271 extratropical cyclones and their frontal systems. The recent severe drought in California is 272 273 coincident with winters in which East Pacific extratropical cyclone activity is much reduced. Both California precipitation and East Pacific cyclone activity display weak decreasing trends 274 between 1979/80 to 2014/15, but neither trend is statistically significant. 275

Examination of climate projections made by 29 CMIP5 models under the high emission 276 RCP8.5 pathway suggests a small ensemble mean increase in precipitation over much of 277 California, consistent with a small projected increase in East Pacific extratropical cyclone 278 activity. Nevertheless, consistent with previous studies, this region lies close to the transition 279 zone between projected mid-latitude increase and subtropical decrease in precipitation, thus there 280 is significant model-to-model variability in the projected change in precipitation, with some 281 models projecting future decrease while a majority of the models (23 out of 29) project increase. 282 Our analyses show that model-to-model differences in projected change in California 283 precipitation is highly correlated with model-to-model differences in projected change in East 284 Pacific extratropical cyclone activity. In fact, using the relationship between interannual cyclone 285 activity and precipitation variability from the historical period, one can accurately "predict" each 286 model's projected California precipitation change using its projected change in East Pacific 287 288 extratropical cyclone activity. All these demonstrate that variability and change in California precipitation is highly modulated by those in East Pacific extratropical cyclone activity. 289

Neelin et al. [2013] showed that model-to-model differences in projected change in 290 291 California precipitation are highly correlated with those of the upper level jet stream, and they hypothesized that this correlation represents the effect of the jet steering storm tracks (or 292 extratropical cyclones) towards the California coast. In this study, we have directly demonstrated 293 this physical link between storm tracks and California precipitation. Comparing our results to 294 those of Neelin et al. [2013], the relationship between projected changes in California 295 precipitation and East Pacific extratropical cyclone activity is apparently stronger than the 296 297 relationship between precipitation and upper level jet (compare our Fig. 4 to their Figs. 7 and 10), which is not surprising given the more direct physical linkage between extratropical 298 cyclones and precipitation. 299

Our analyses of results taken from 10 ensemble members of CSIRO-Mk3.6.0 suggest that part of the model-to-model differences found in this study may be due to internal climate variability, but uncertainties in model response likely still account for much of this spread. The physical mechanisms that give rise to these differences in model response in extratropical cyclone activity and its associated precipitation thus provide a research target of leading importance for California water resources.

Acknowledgments. The authors would like to thank the Earth System Grid and the climate
 modelling centers for making available the CMIP5 model data. The Stony Brook authors are
 supported by NSF grant AGS-1261311, and JDN by NSF grant AGS-1102838.

- 310
- 311

312 **References**

- Adler, R.F., G.J. Huffman, A. Chang, R. Ferraro, P. Xie, J. Janowiak, B. Rudolf, U. Schneider,
 S. Curtis, D. Bolvin, A. Gruber, J. Susskind, and P. Arkin (2003), The Version 2 Global
 Precipitation Climatology Project (GPCP) Monthly Precipitation Analysis (1979Present), J. Hydrometeor., 4, 1147-1167.
- AghaKouchak, A., L. Cheng, O. Mazdiyasni, and A. Farahmand (2015), Global warming and
 changes in risk of concurrent climate extremes: Insights from the 2014 California
 drought. *Geophys. Res. Lett., in press*, doi: 10.1002/2014GL062308.
- Bengtsson, L., K. I. Hodges, and N. Keenlyside (2009), Will extratropical storms intensify in a
 warmer climate? *J. Climate*, 22, 2276-2301.
- Cayan, D. R., K. T. Redmond, and L. G. Riddle (1999), ENSO and hydrologic extremes in the
 Western United States. J. Climate, 12, 2881-2893.
- Cayan, D. R., E. P. Mauer, M. D. Dettinger, M. Tyree, and K. Hayhoe (2008), Climate change
 scenarios for the California region. *Climatic Change*, 87, S21-S42.
- Chang, E. K. M. (2013), CMIP5 projection of significant reduction in extratropical cyclone
 activity over North America, *J. Clim.*, 26, 9903-9922.
- Chang, E. K. M., and S. Song (2006), The seasonal cycles in the distribution of precipitation
 around cyclones in the Western North Pacific and Atlantic, *J. Atmos. Sci.*, 63, 815-839.
- Chang, E. K. M., S. Lee, and K. L. Swanson (2002), Storm track dynamics, *J. Clim.*, *15*, 2163-2183.
- Chang, E. K. M., Y. Guo, and X. Xia (2012), CMIP5 multimodel ensemble projection of storm
 track change under global warming. *J. Geophys. Res.*, *117*, D23118.
- Collins, M., R. Knutti, J. Arblaster, J.-L. Dufresne, T. Fichefet, P. Friedlingstein, X. Gao, W.J. 334 Gutowski, T. Johns, G. Krinner, M. Shongwe, C. Tebaldi, A.J. Weaver and M. Wehner, 335 (2013), Long-term Climate Change: Projections, Commitments and Irreversibility. In: 336 337 Climate Change 2013: The Physical Science Basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Stocker, 338 T.F., D. Qin, G.-K. Plattner, M. Tignor, S.K. Allen, J. Boschung, A. Nauels, Y. Xia, V. 339 Bex and P.M. Midgley (eds.)]. Cambridge University Press, Cambridge, United Kingdom 340 and New York, NY, USA. 341
- Deser, C., A. Phillips, V. Bourdette, and H. Teng (2012), Uncertainty in climate change projections: the role of internal variability, *Clim. Dyn.*, *38*, 527-546.
- Dettinger, M. D., D. R. Cayan, H. F. Diaz, and D. M. Meko (1998), North-south precipitation
 patterns in Western North America on interannual-to-decadal timescales. *J. Climate*, 11, 3095-3111.
- Griffin, D., and K.J. Anchukaitis (2015), How unusual is the 2012-2014 California drought?
 Geophys. Res. Lett., in press, doi:10.1002/2014GL062433.
- Hawkins, E., and R. Sutton (2011), The potential to narrow uncertainty in projections of regional
 precipitation change, *Clim. Dyn.*, *37*, 407-418.

- Held, I. M., S. W. Lyons, and S. Nigam (1989), Transients and the extratropical response to El
 Niño. J. Atmos. Sci., 46, 163-174.
- Hoskins, B. J., and K. I. Hodges (2002), New perspectives on the Northern Hemisphere winter
 storm tracks, *J. Atmos. Sci.*, *59*, 1041-1061.
- Kalnay, E., and coauthors (1996), The NCEP/NCAR 40-year reanalysis project, *Bull. Amer. Meteor. Soc.*, 77, 437-470.
- Knutti, R., R. Furrer, C. Tebaldi, J. Cermak, and G. A. Meehl (2010), Challenges in combining
 projections from multiple climate models. J. Clim., 23, 2739-2758.
- Knutti, R., D. Masson, and A. Gettelman (2013), Climate model genealogy: Generation CMIP5
 and how we got there, *Geophys. Res. Lett.*, 40, 1194-1199.
- Lau, N.-C. (1978), On the three-dimensional structure of the observed transient eddy statistics of
 the Northern Hemisphere wintertime circulation. J. Atmos. Sci., 35, 1900-1923.
- Maloney, E. D. and coauthors (2014), North American Climate in CMIP5 Experiments. Part III:
 Assessment of 21st Century Projections, *J. Climate*, 27, 2230-2270, doi:10.1175/JCLI-D 13-00273.1.
- Meehl, G., T. F. Stocker, W. Collins, and coauthors (2007), Global climate projections. Climate
 Change 2007: The Physical Science Basis. Contribution of Working Group I to the
 Fourth Assessment Report of the Intergovernmental Panel on Climate Change, S.
 Solomon, D. Qin, and M. Manning, Eds., Cambridge University Press, Cambridge and
 New York, 996pp.
- Neelin, J. D., M. Munnich, H. Su, J. E. Meyerson, and C. E. Holloway (2006), Tropical drying
 trends in global warming models and observations, *Proc. Natl. Acad. Sci.*, 103, 6110 6115.
- Neelin, J. D., B. Langenbrunner, J. E. Meyerson, A. Hall, and N. Berg (2013), California winter
 precipitation change under global warming in the Coupled Model Intercomparison
 Project Phase 5 ensemble, *J. Clim.*, 26, 6238-6256.
- Pierce, D. W., T. Das, D. R. Cayan, E. P. Mauer, N. L. Miller, Y. Bao, M. Kanamitsu, K.
 Yoshimura, M. A. Snyder, L. C. Sloan, G. Franco, and M. Tyree (2012), Probabilistic
 estimates of future changes in California temperature and precipitation using statistical
 and dynamical downscaling, *Climate. Dyn.*, doi: 10.1007/s00382-012-1337-9.
- Seager, R., M. Hoerling, S. Schubert, H. Wang, B. Lyon, A. Kumar, J. Nakamura, and N.
 Henderson (2014a), Causes and predictability of the 2011-14 California drought.
 DTF/NIDIS Assessment Report, doi:10.7289/V58K771F
- Seager, R., J. D. Neelin, I. Simpson, H. Liu, N. Henderson, T. Shaw, Y. Kushnir, M. Ting, and
 B. Cook (2014b), Dynamical and thermodynamical causes of large-scale changes in the
 hydrological cycle over North America in response to global warming, *J. Climate*, 27,
 7921–7948, doi:10.1175/JCLI-D-14-00153.1.
- Straus, D. M., and J. Shukla (1997), Variations of midlatitude transient dynamics associated with
 ENSO. J. Atmos. Sci., 54, 777-790.
- Tebaldi, C., and J. M. Arblaster (2014), Pattern scaling: Its strengths and limitations, and an
 update on the latest model simulations. *Climatic Change*, 122, 459-471,
 doi:10.1007/s10584-013-10.
- Wallace, J. M., G.-H. Lim, and M. L. Blackmon (1988), Relationship between cyclone tracks,
 anticyclone tracks and baroclinic waveguides, *J. Atmos. Sci.*, 45, 439-462.
- 395



Fig. 1: a) NCEP DJF bandpass-filtered SLP variance, *pp*, climatology from 1979/80 to 2014/15.
b) CMIP5 29-model ensemble mean historical pp climatology from 1961 to 2000. c) Correlation
map of GPCP DJF precipitation averaged over the CA precipitation box and NCEP DJF pp at
each grid point from 1979/80 to 2014/15. The red contour shows the correlation is at 95%
statistical significant level. d) Mean of correlation maps of DJF precipitation in CA box and DJF
pp from 1961 to 2000. e) The shaded region is used to calculate precipitation in CA. The dashed
box shown in panels a-d is the box used to compute East Pacific extratropical cyclone activity.



405

406

Fig. 2: a) Scatterplot of NCEP reanalysis pp (hPa²) in the East Pacific extratropical cyclone activity box (see Fig. 1c) vs. GPCP precipitation (mm/day) in California region (see Fig. 1e) during DJF from 1979/80 to 2014/15. b) The time series of the two quantities. Precipitation is in blue color and has a decreasing trend of -0.025 mm/day/year, and pp is in red color and has a decreasing trend of -0.136 hPa²/year. Neither precipitation nor pp shows a statistically significant trend.



Fig. 3: a) CMIP5 multi-model ensemble mean projection of DJF precipitation (mm/day) change 416 417 from 1961-2000 to 2059-2098. The red line shows the 95% inter-model statistical significant level based on a student's t-test. b) Similar to a) but showing the model agreement. The warm 418 colors show the number of models that project precipitation decrease and cold colors show the 419 number of models that project precipitation increase. c)-d) Same as a)-b) but for projected pp 420 (hPa^2) change. 421



b) pp change * slope vs Pr change

Fig. 4: a) Scatterplot showing CMIP5 East-Pacific pp (hPa²) cyclone activity index change vs 424 precipitation (mm/day) change for each model. Both pp and precipitation change are the 425 difference between RCP8.5 run from 2059-2098 DJF and historical run from 1961-2000 DJF. 426 Black line shows the regression line (and correlation, corr, and slope values) corresponding to all 427 CMIP5 models. Blue line shows the regression line (and *corr and *slope values) corresponding 428 to when MRI-CGCM3 and MRI-ESM1 are excluded. b) Ordinate shows model projected 429 California DJF precipitation change (2059-2098 minus 1961-2000), and abscissa shows the DJF 430 pp change (2059-2098 minus 1961-2000) multiplied by slope for precipitation change per unit pp 431 change derived from historical runs in each model (see Supplemental Fig. S1). MME Mean 432 (black) point shows the multi-model ensemble mean of the value in x-axis and y-axis. Black line, 433 434 corr and slope values correspond to all CMIP5 models. Blue line and *corr and *slope values correspond to when MRI-CGCM3 and MRI-ESM1 are excluded. 435