

Supplemental Material

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Supplemental Material for "Skillful climate forecasts of the tropical Indo-Pacific Ocean using model analogs"

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Perfect-model skill of analog ensemble mean forecast, at 24-month lead



Figure S1 Perfect-model skill of ensemble-mean model-analog forecast of SSTA at 24-month 16

17 lead for the verification period in (a) CM2.1 (b) CM2.5 FLOR (c) CCSM4 and (d) CESM1.

18 Shading denotes local anomaly correlation and contours RMS skill score. For each model, the

19 first two hundred years of its control run was used as the verification period, with the remaining 20

data used for the data library. Contour/shading interval is 0.05/0.1.

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26 Figure S2 Perfect-model skill of ensemble-mean model-analog forecast of SSTA at 6-month lead 27 for the verification period in (a) CM2.1 (b) CM2.5 FLOR (c) CCSM4 and (d) CESM1. Shown is $\rho^2 + \epsilon^2$, where ρ and ϵ denote local correlation and standardized RMS error. For each model, 28

29 the first two hundred years of its control run was used as the verification period, with the

30 remaining data used for the data library.



31 32 condition to 24-month lead. Shown is $\rho^2 + \epsilon^2$, where ρ and ϵ denote correlation and 33

34 standardized RMS error, taking into account all grid points in the training region and all forecasts

35 with identical lead time.

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42 43 Figure S5. Comparison of month 6 forecasts of Nino3.4 SST anomalies between analog forecasts 44 and corresponding NMME forecasts in (a) CM2.1 (b) CM2.5 FLOR (c) CCSM4 (d) CESM1 and (e) four-model grand mean. In the comparison, only ensemble means are employed. The count 45 46 increases by 1 when RMS skill score of analog forecasts is more than that of corresponding NMME forecasts, and decreases by 1 otherwise. The count is accumulated forward in time for 47 48 each model separately, over all initial months and years, thereby tracing out a random walk. The 49 red and blue lines indicate the range of counts that would be obtained 95% of the time under 50 independent Bernoulli trials for p = 1/2. A random walk extending above the red lines indicates 51 that analog forecasts display higher RMS skill score significantly more often than expected for independent Bernoulli trials (i.e., analog forecasts are more skillful than the corresponding 52 53 NMME forecasts). On the other hand, a random walk extending below the blue line indicates 54 that analog forecasts display higher skills significantly less often than expected for independent Bernoulli trials (i.e., analog forecasts are less skillful than the corresponding NMME forecasts). 55 56 Readers are referred to Delsole and Tippett (2016) for more details on comparing forecast skills 57 based on random walks. 58



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Figure S6. Analog hindcast skill of observed SST variations at six-month lead based on (a) CM2.1 (b) CM2.5 FLOR (c) CCSM4 and (d) CESM1. Shadings denote correlation. Here, analogs are based on a distance taking into account velocity in trajectory: $d^2(t, t') =$

 $\sum_{i=1}^{2} \sum_{j=1}^{J} \left(\left(\frac{x_j^i(t)}{\sigma_x^i} - \frac{y_j^i(t')}{\sigma_y^i} \right)^2 + \left(\frac{dx_j^i(t)}{\sigma_x^i} - \frac{dy_j^i(t')}{\sigma_y^i} \right)^2 \right) \text{, where } dx_j^i(t) = x_j^i(t) - x_j^i(t-1) \text{ and}$ $dy_j^i(t') = y_j^i(t') - y_j^i(t'-1). \text{ Compare with the upper four panels of the left hand side of Fig.}$ 10.