**ENSO多模式实时预测系统**

以自然资源部第二海洋研究所卫星海洋环境动力学国家重点实验室作为依托单位，由陈大可院士任首席的国家自然科学基金委重大项目“ENSO的变异机理和可预测性研究”（41690120）第四课题研发的ENSO多模式集合预报系统研发取得重要进展，目前系统已完成测试、评价，并开展实时预报。

该多模式系统包含项目参与单位海洋二所陈大可团队、中科院大气所段晚锁团队和中科院海洋所张荣华团队发展的5个不同复杂程度的海气耦合模式（表1），各模式分别采用集合卡尔曼滤波（EnKF）、非线性强迫奇异向量同化（NFSV-assimilation; Tao et al., 2019）、Nudging同化以及线性插值方法同化大气、海洋资料进行模式初始化。利用随机最优扰动（Stochastic optimals，简称SOs，Kleeman and Moore 1997）、气候相关的奇异向量（Climatically relevant singular vectors，简称CSVs，Kleeman et al. 2003；Tang et al. 2006）和正交条件非线性最优扰动（Orthogonal Conditional Nonlinear Optimal Perturbations，简称O-CNOPs，Duan and Huo 2016）的集合方法度量初始条件或模式误差的不确定性，并产生集合成员。过去一个多世纪的回报试验结果表明：该系统Niño3.4指数的预报技巧已进入国际ENSO预测水平的第一梯队，提前6个月的预报技巧能够达到0.83，提前12个月能够达到0.59，系统能够提前12个月提供有效的ENSO预报。

该系统已与2020年9月开始定期开展实时预报，2022年10月的预报结果显示，未来热带太平洋地区拉尼娜状态将维持（图1）。

表1 多模式ENSO集合系统模式信息一览表

|  |  |  |  |
| --- | --- | --- | --- |
| **参与单位** | **模式** | **同化方法** | **集合方法** |
| 海洋二所 | SIO\_ICM(Tang et al. 2018;Liu et al. 2019;Gao et al. 2020) | EnKF | EnKF+SOs |
| 海洋二所 | SIO\_TICM(Song et al. 2018) | Nudging | / |
| 海洋二所 | SIO\_CESM(Liu et al. 2020;Song et al. 2020) | Nudging | CSVs |
| 中科院大气所 | IAP\_NFSV-ICM(Duan and Huo 2016;Tao et al. 2020) | Nudging和NFSV-assimilation | O-CNOPs |
| 中科院海洋所 | IOCAS-ICM(Zhang and Gao 2016a,b) | 线性插值 | / |



图1 多模式ENSO集合系统2022年10月起报的确定性预报结果

**免责申明：**

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