

Forecasting in the face of ecological complexity:

Number and strength of species interactions determine forecast skill in ecological communities

Uriah Daugaard¹, Stephan B. Munch², David Inauen¹, Frank Pennekamp¹ & Owen L. Petchey¹

¹Department of Evolutionary Biology and Environmental Studies, University of Zurich; ²Department of Ecology and Evolutionary Biology, University of California, Santa Cruz

BACKGROUND

- It is challenging to **forecast** the ecological consequences of climate change, biodiversity loss, etc.
- The **complexity of ecological systems** might render ecology unpredictable¹
- Studies both **support**² and **refute**³ that **complexity hinders forecast skill**
- **Complexity within a system** varies as well: species differ in **how many interactions** they have and **how strong** these are
- Yet, little is known about how **within system complexity** is related to **forecast skill**

RESEARCH QUESTIONS

- Does the **forecasting** of species **abundances** depend:
 1. On the **number of interactions** a species has?
 2. On the **mean strength** of these interactions?
 → Forecast skill vs. complexity **within** system
- Does an **increase in complexity** decrease how well we can **forecast species abundances**?
 - Forecast skill vs. complexity **across** systems

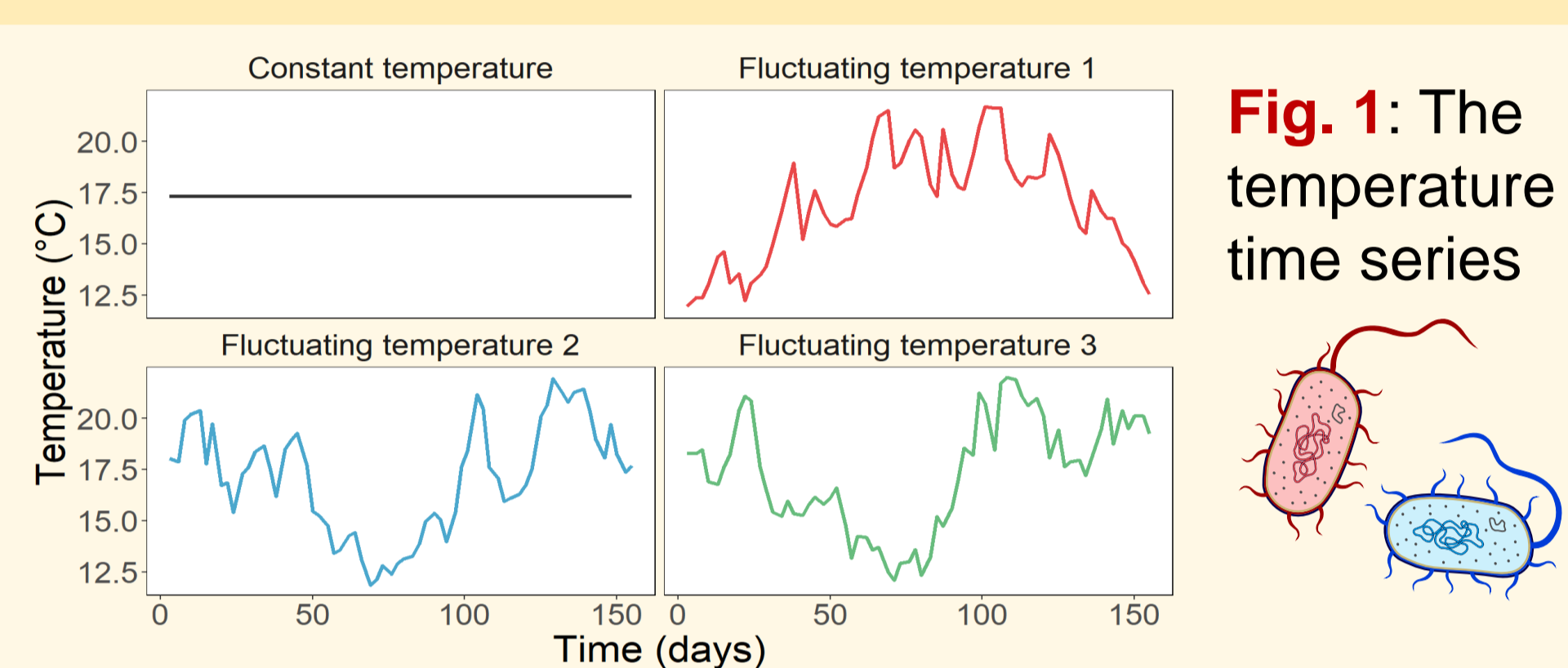
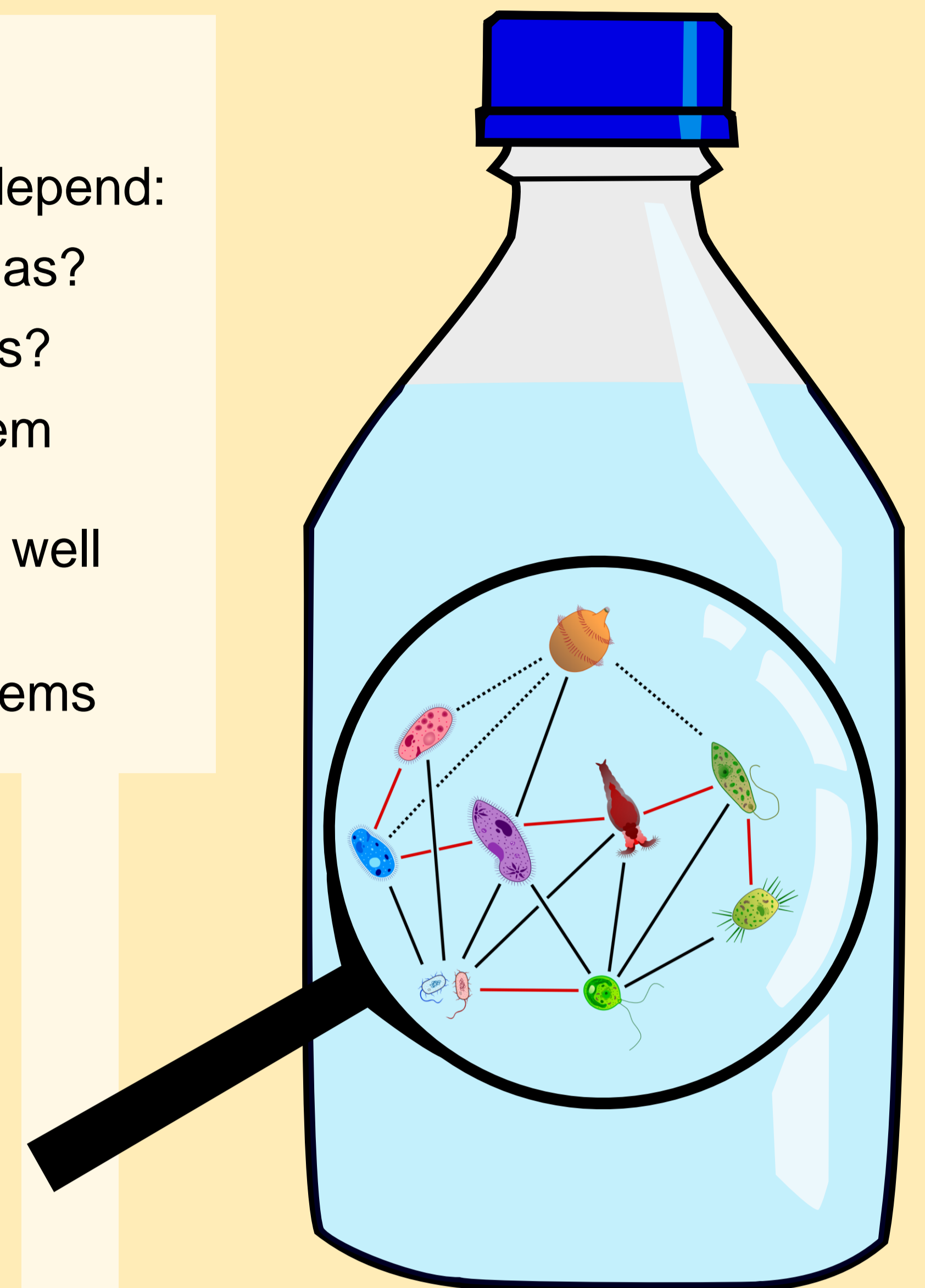


Fig. 1: The temperature time series



METHODS

- Experiment**
- **Microcosms:** tri-trophic microbial communities in 2L **bottles**
 - **Treatment:** 1 constant and 3 fluctuating temperatures (Fig. 1)
 - The fluctuating temperatures **complexify** the system
 - **Replicates:** 9 at constant and 3 at each fluctuating temperature
 - **Sampling:** 3 times per week (Mo, We, Fr) for 22 weeks

- Analyses**
- **Time series:** detrended and standardized
 - **Forecast error of abundances** determined with three methods: EDM, ARIMA & RNN
 - **Number of interactions:** estimated with CCM EDM
 - **Interaction strengths:** estimated with S-map EDM & MARSS

RESULTS

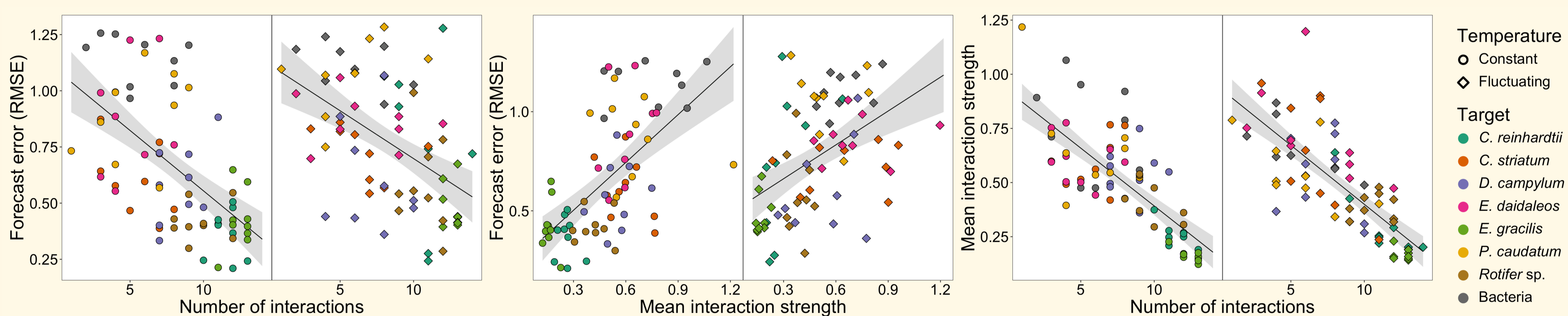


Fig. 2: The more interactions a species had, the weaker these interactions were and the better its abundance was predicted

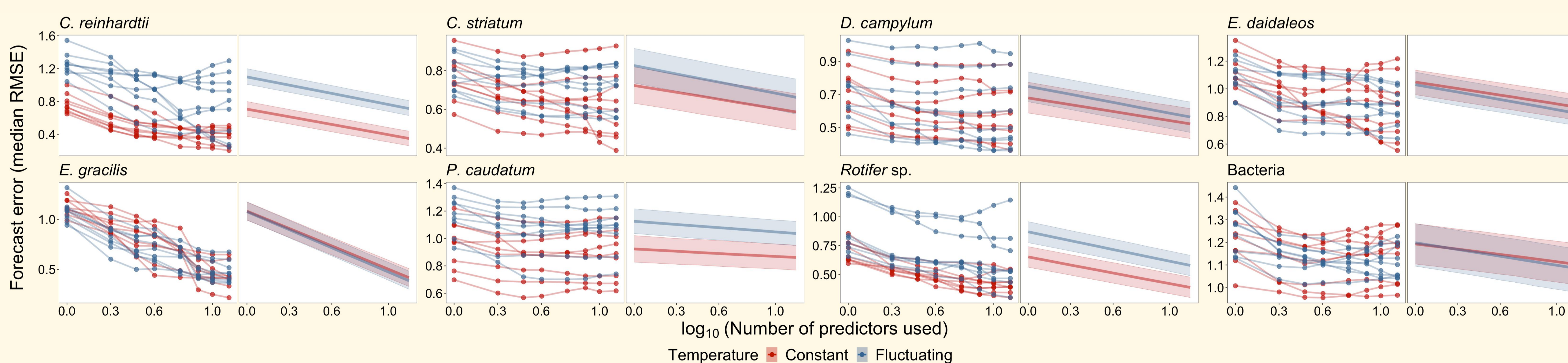


Fig. 3: The greater system complexity (i.e. the fluctuating temperatures) increased the forecast error for 3 out of 8 species

CONCLUSIONS

Within the system, complexity improved forecasting (Fig. 2)

- To forecast abundances equally well different amounts of data are required for different species
- Increased system complexity decreased the forecast skill of some species (Fig. 3)
- The effect of complexity can be species-specific and of different sign within and across systems

REFERENCES

1. Beckage, B., Gross, L. J., and Kauffman, S. (2011). The limits to prediction in ecological systems. *Ecosphere*
2. Jonsson, T., Kaartinen, R., Jonsson, M., et al. (2018). Predictive power of food web models based on body size decreases with trophic complexity. *Ecol. Lett.*
3. Mougi, A. (2017). Spatial complexity enhances predictability in food webs. *Sci. Rep.*

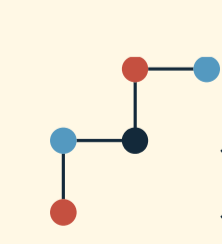
GLOSSARY

- ARIMA:** Auto-Regressive Integrated Moving Average
- CCM:** Convergent Cross Mapping
- EDM:** Empirical Dynamic Modeling
- MARSS:** Multivariate Auto-Regressive State Space
- RNN:** Recurrent Neural Network
- S-map:** Sequential Mapping

ACKNOWLEDGEMENTS



University of Zurich



Swiss National Science Foundation

See the publication in Ecology Letters!

CONTACT INFORMATION

@DaugaardUriah

uriah.daugaard@uzh.ch

