

面向昆蒙框架的生物多样性快速评估和早期预警

俞乐

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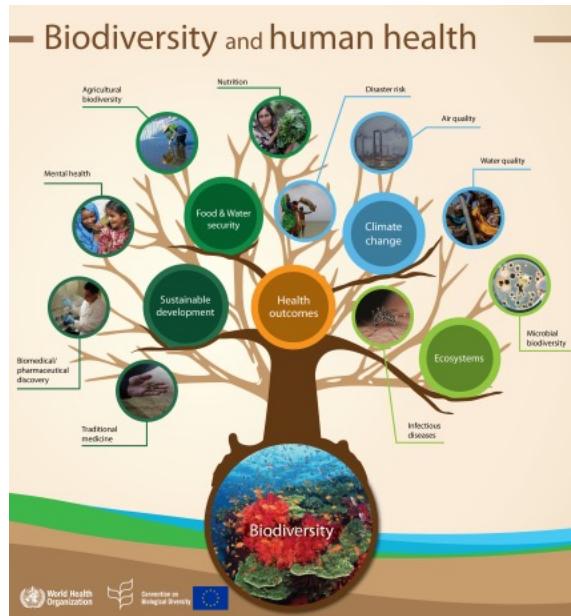
Kunming-Montreal

GLOBAL BIODIVERSITY FRAMEWORK

人类的生存发展需要生物多样性

生物多样性(生态系统、物种、遗传多样性)并非“锦上添花”，而是人类生存的基石

- 粮食安全：全球约**75%**的粮食作物依赖动物传粉，包括蜜蜂、蝴蝶、鸟类等 (IPBES, 2016)
- 气候调节：热带森林每年吸收约11.2Gt CO₂，占全球人类排放的**30%** (Friedlingstein et al., 2023)
- 渔业生计：全球渔业产业为约**6,000万人**提供就业，每年提供超**1.5亿吨**水产品 (FAO, 2022)
- 药用资源：超过**50%**的现代药物源自自然物种，如**紫杉醇** (抗癌) 来自红豆杉，**青蒿素** (抗疟疾) 来自青蒿 (WHO, 2015)



(WHO, CBD, EU, 2015)



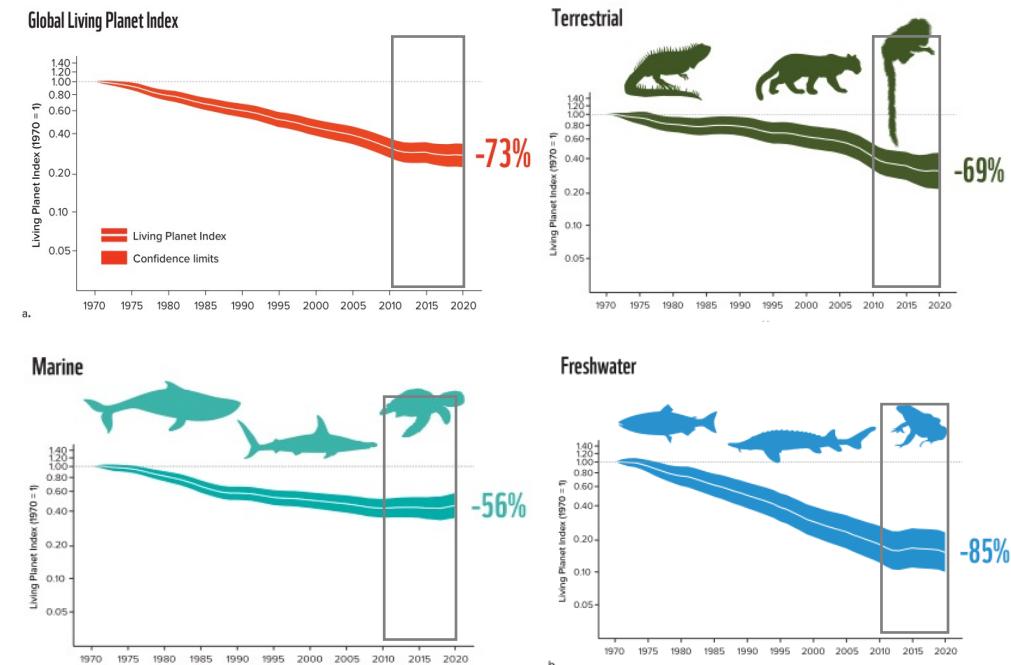
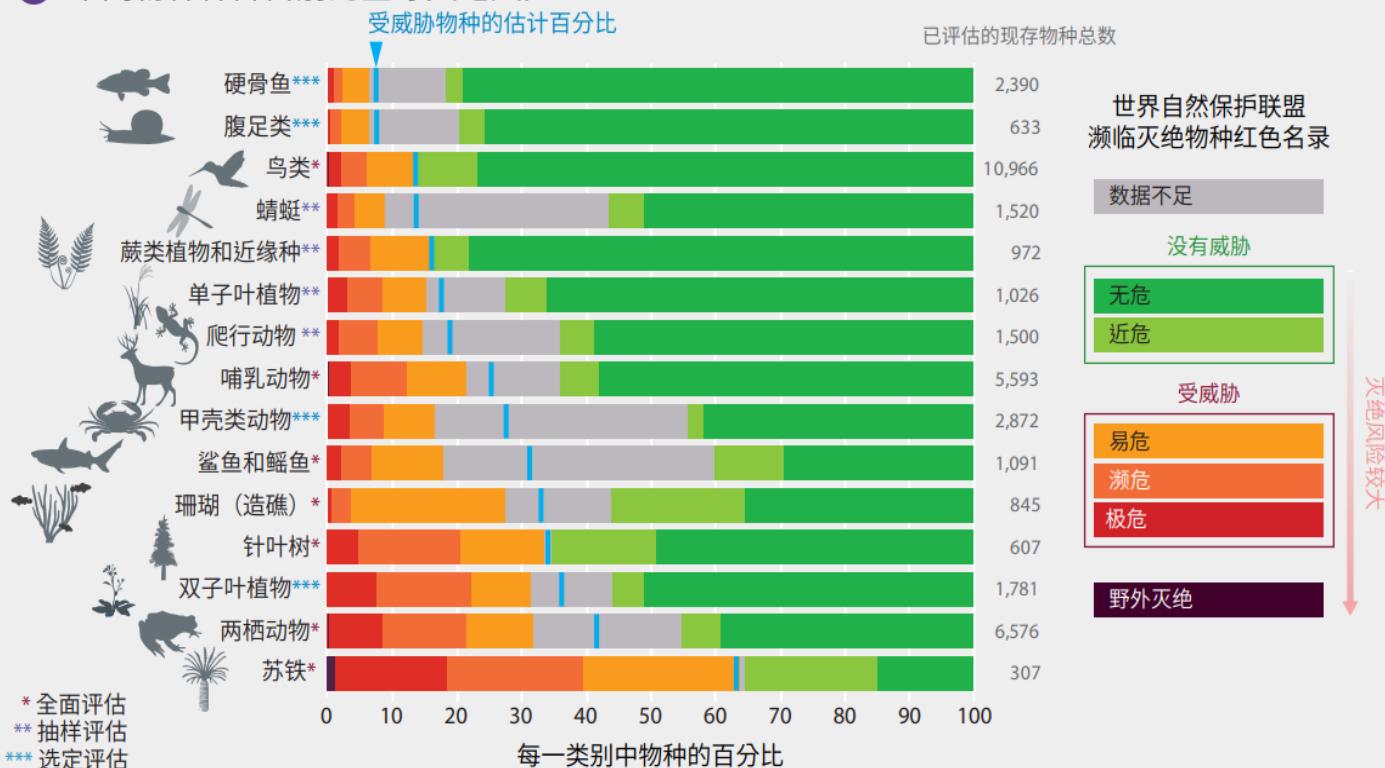
(WWF, 2024)



全球生物多样性丧失现状

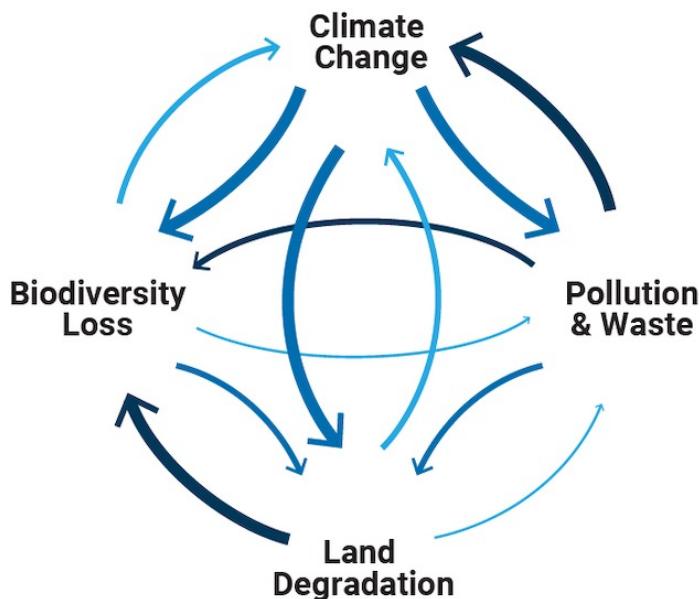
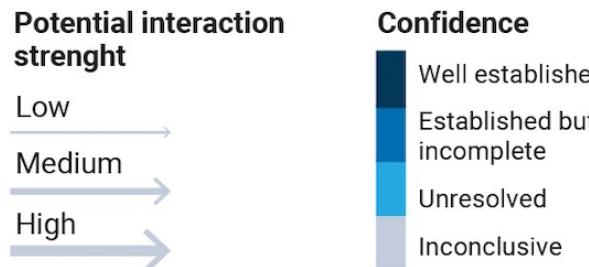
- 人类活动造成的栖息地丧失、气候变化、环境污染等驱动因素正持续威胁地球的生物多样性
- 全球物种灭绝速度比背景灭绝速度(过去一千万年的平均速度)高至少几十倍到几百倍，而且仍在加速
- 目前面临灭绝威胁的物种的平均比例约为25%
- 地球生命力指自1970年以来下降73%

A 不同物种群目前的全球灭绝风险

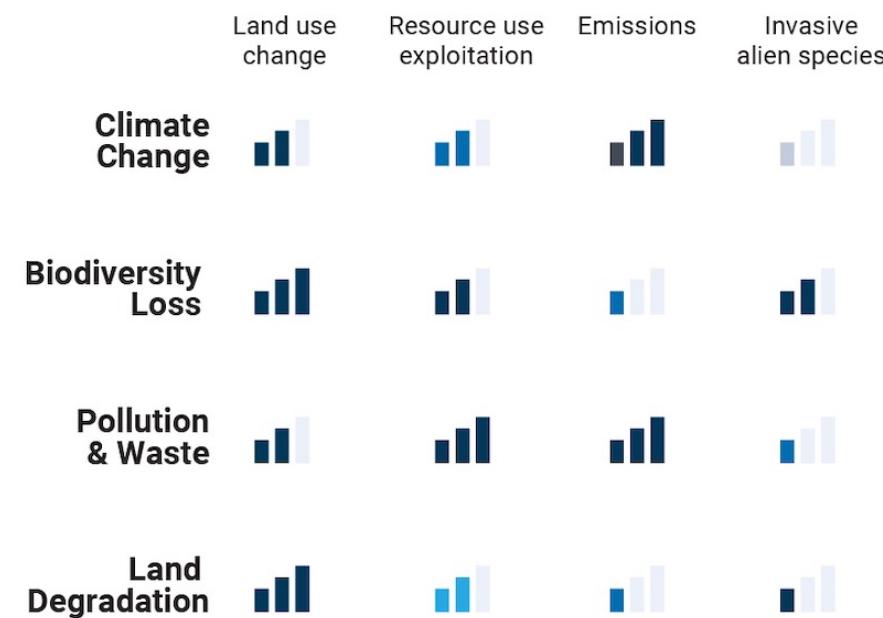
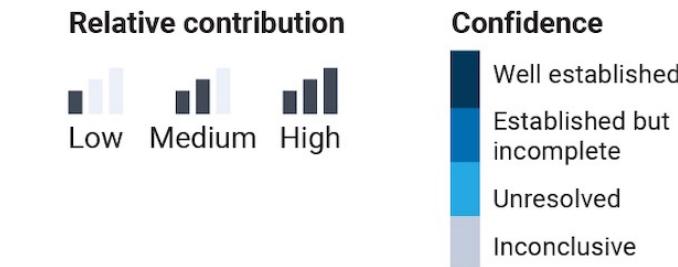


全球环境危机互相交织

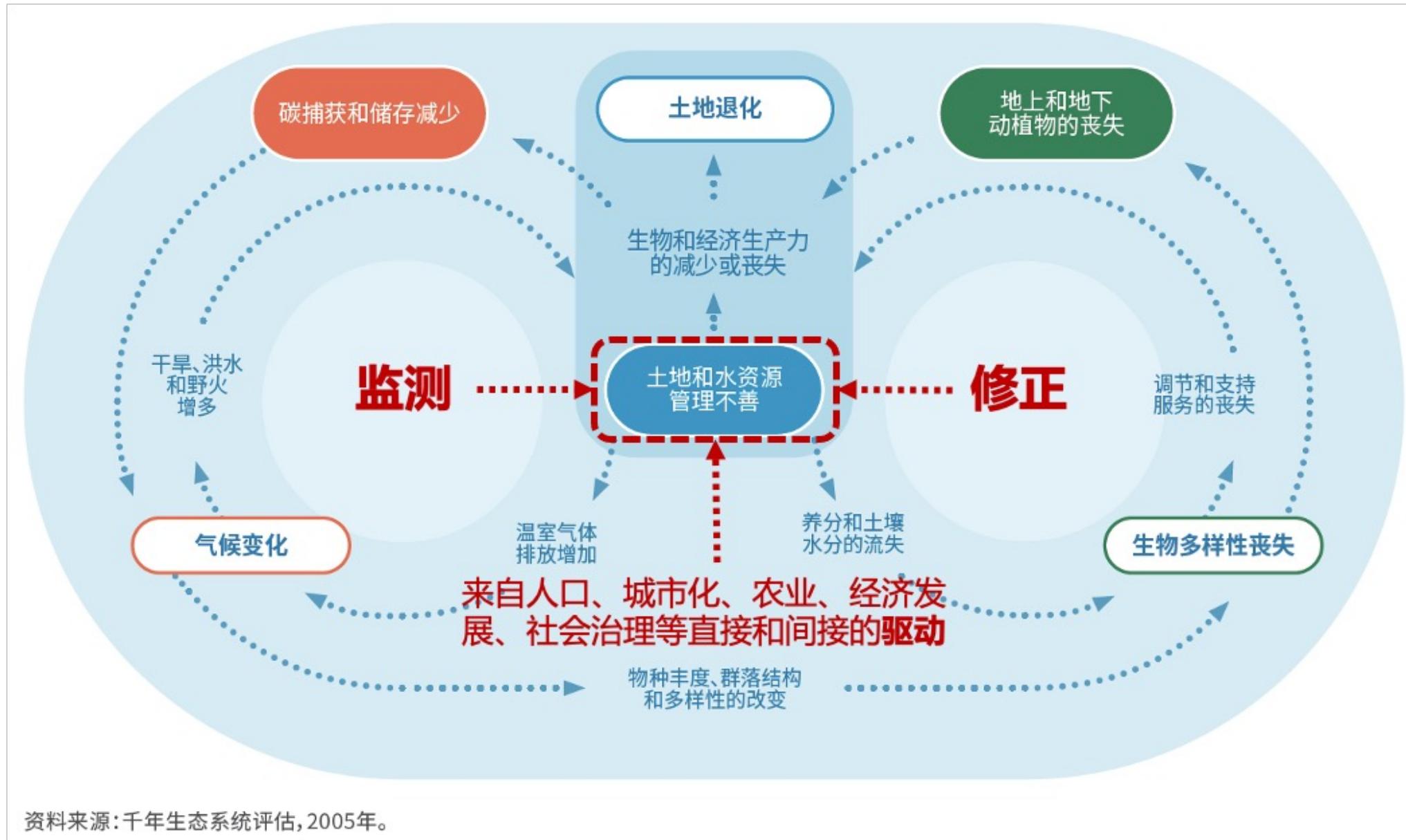
a) Potential interaction strength between each crisis



b) Relative contribution of the different pressures to each crisis

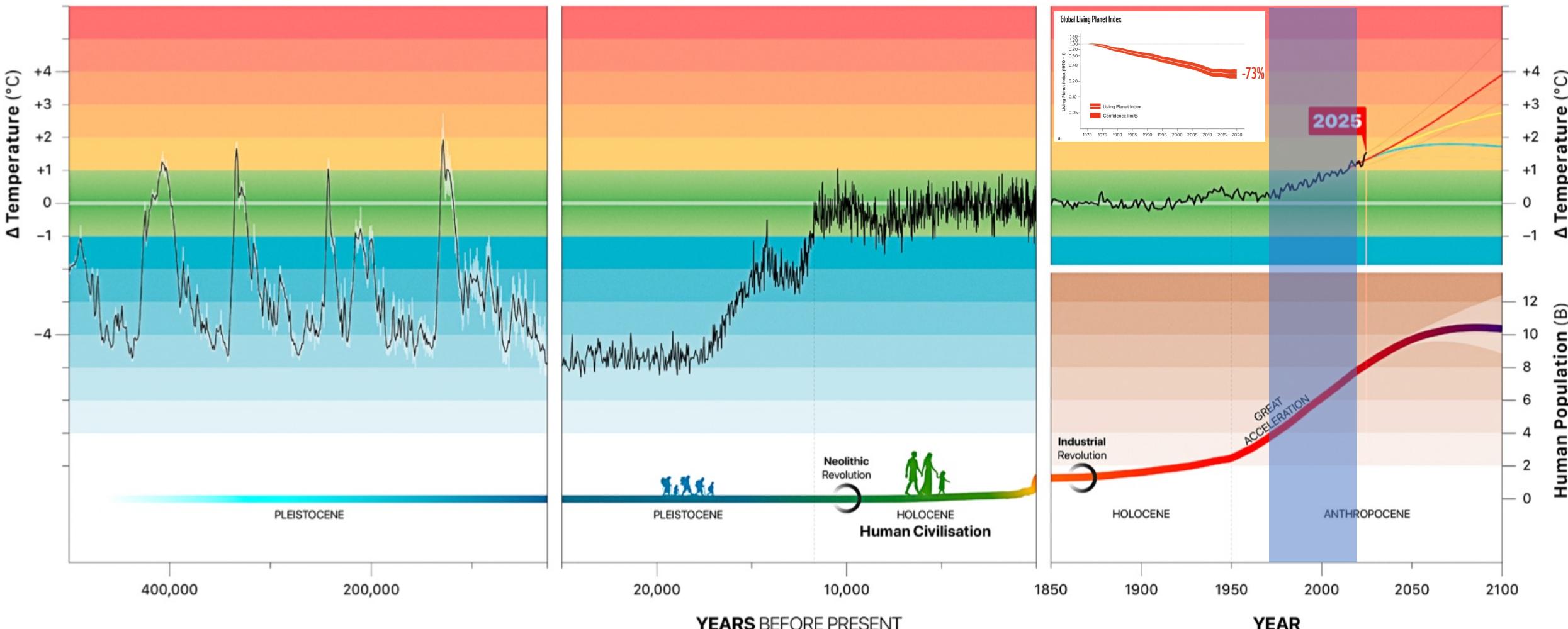


及时监测与评估



快速评估和早期预警

“一万多年来，人类一直生活在气候极为稳定的时期（即“绿色走廊”），并在这一时期演化发展，不断调整自身的技术与文化。然而，随着包括气候变化在内的多项行星边界被突破，这一稳定时期已然终结，我们正步入一个全新而危险的阶段——在这一阶段中，仍在持续增长的全球人口必须全力保障人类福祉。”



信息技术+AI

a 缺陷需求

林奈缺陷
分类命名和描述停滞不前

普雷斯顿缺陷
无法准确检测种群中真实丰度的变化

华莱士缺陷
分类学和区域性信息分布存在偏差

哈钦森缺陷
难以理解生物对环境的真实响应

朗克尔缺陷
缺乏关于生物形态和功能的数据

达尔文型短板
无法确定系统发育树的抽样难题

艾尔顿型短板
关于物种相互作用的数据非常有限

b AI 的机遇

林奈缺陷
通过自动化传感器发现未鉴定的分类单元

普雷斯顿缺陷
在传感器数据中识别个体

华莱士缺陷
从多模态数据流中建模物种分布

哈钦森缺陷
从卫星影像中推断栖息地，或通过性状推断行为

朗克尔缺陷
通过图像或文本挖掘提取形态学特征

达尔文缺陷
实现高通量系统发育拼接

艾尔顿缺陷
检测和建模相关的物种相互作用

对应研究领域

所有领域

- 种群生态学
- 群落生态学
- 生物地理学
- 生态系统生态学

- 功能生态学
- 进化学
- 食物网生态
- 授粉生态

保护实践示例

- 濒危种群
- 红色名录指数
- 生命星球指数
- “30x30”保护规划

- 临界点
- 生态系统韧性
- 关键物种
- 生态系统红色名录
- 保护等级（如 EDGE）

a AI 的当前应用

初级数据采集

红外相机
声学监测
卫星影像

改进的数据填补方法

分析

生态建模

知识

生态应用
保护应用

b AI 的未来应用

初级数据采集

环境DNA
声学监测
红外相机
卫星影像

快速提取已有信息

改进的数据填补方法

分析

- 自动综合
- 元分析
- 生态建模
- 假设生成

优化数据到应用的流程

知识

生态应用
保护应用

AI 可协助完成的任务

稀有事件和异常检测

数据处理效率提升

更好的统计推断与预测

提升研究效率（如优化抽样策略的使用）

发现潜在模式

跨物种到群落层级的分类、空间和时间尺度连接

快速信息提取

• 在大型现有数据集或灰色文献中查询和提取图像信息
• 查询和总结文本文件（大型语言模型，LLMs）（例如：濒危物种评估）

初级数据采集

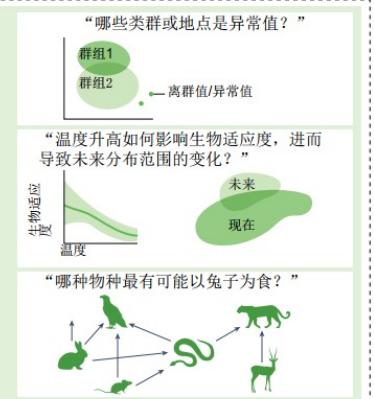
探索与假设生成

• 持续改进传感器数据的监督分类（如红外相机、声学监测、无人机、卫星影像）
• 长尾分类
• 模型泛化
• 多模态AI
• 异常检测
• 边缘AI（Edge AI）

生态建模

• 在不同组织层级之间进行建模，从物种到群落
• 用于弥补数据缺口的数据填补
• 使用知识引导的机器学习
• 因果建模
• 可解释的AI

模型可用于理论生态学问题，例如适应度如何随温度变化，并可用于跨时间或空间的预测，或缺数据的填补。



研究现状

1 长时序生物多样性遥感监测

进展：多尺度、全方位的监测能力提升，以及新技术如无人机和激光雷达的应用，使得生物多样性信息的获取更加直接和精确

局限：

- 多源遥感数据的海量性和复杂性，给**数据处理、运算能力、存储空间**带来挑战
- **空间分辨率**的限制导致微观生态特征的捕捉能力欠佳
- 敏感性不足，面临气候变化和极端事件时的**快速响应能力**亟需增强

传统生物多样性监测

范围较小且耗时耗力，受限于指示物种



(Alberta Biodiversity Monitoring Institute)

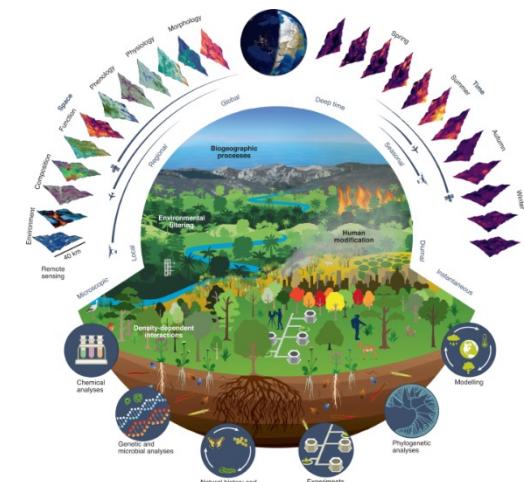
生物多样性关键变量
EBVs, Essential
Biodiversity Variables

生态系统 — 物种 — 基因

气候变化关键变量
ECVs, Essential
Climate Variables

基于遥感的生物多样性监测

大范围长时间序列综合自动观测，直接有效

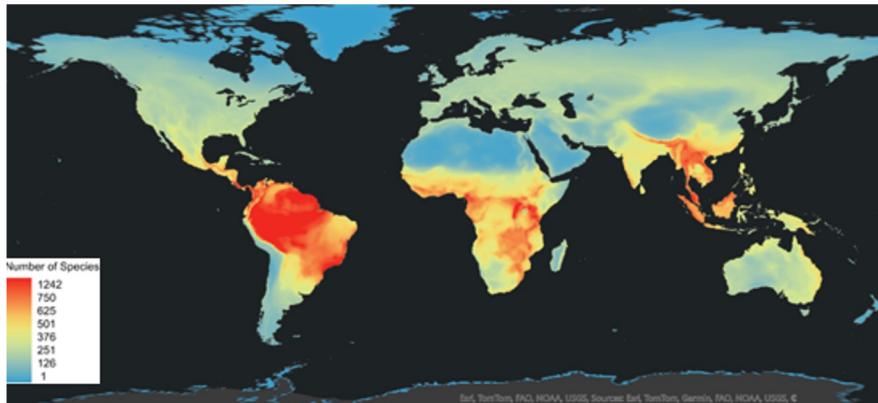


(Cavender-Bares, J., *Nature Ecology & Evolution*, 2022)

研究现状

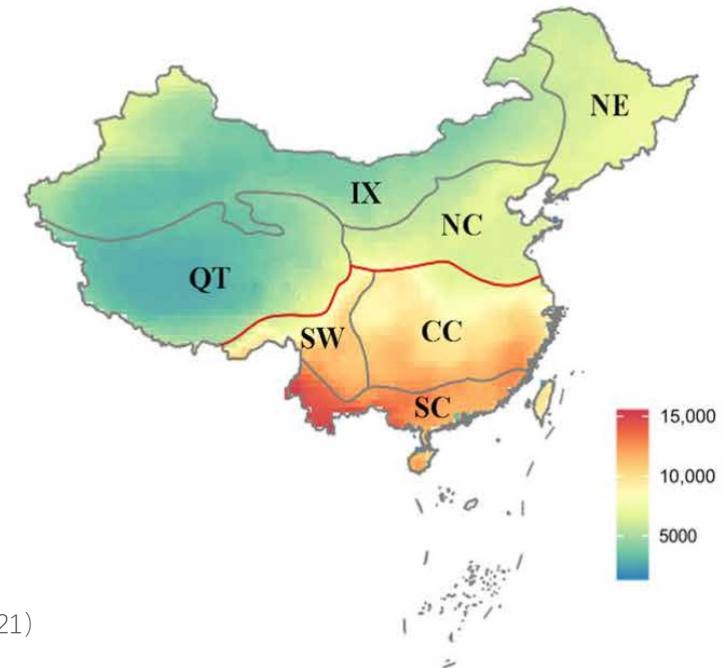
2 生物多样性制图

进展：遥感和GIS技术使全球物种多样性制图取得显著进展，系统发育树和高通量测序技术推动了系统发育和基因多样性研究



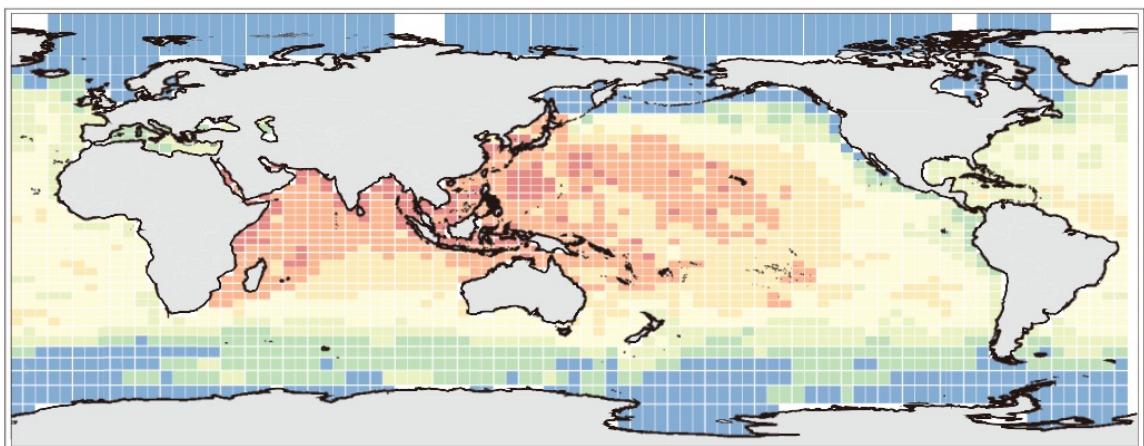
全球陆生脊椎动物物种多样性

IUCN Red List 2024)



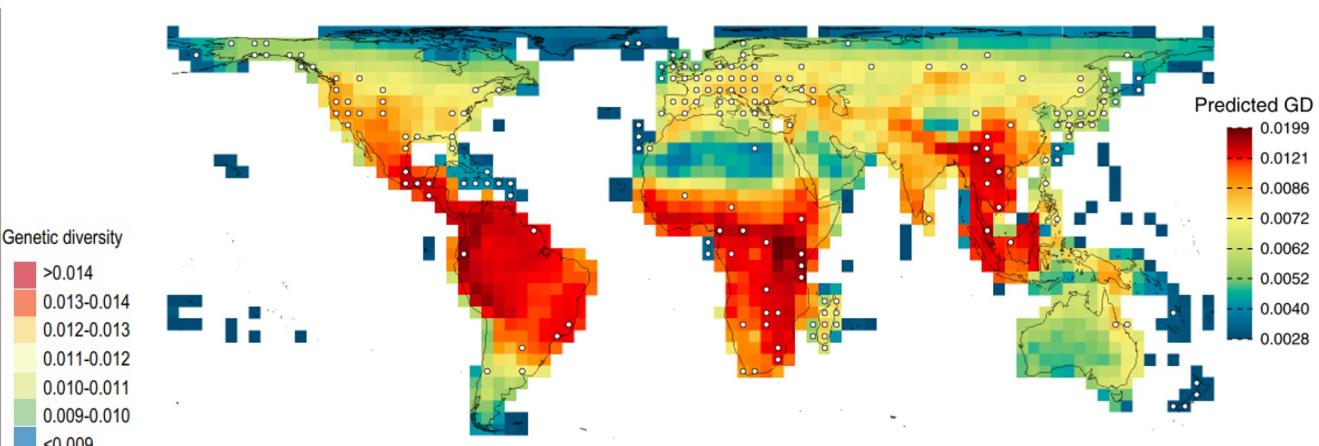
中国陆生脊椎动物系统发育多样性

(Hu, et al., *Science Advances*, 2021)



全球海洋物种遗传多样性空间格局

(Fan et al., *National Science Review*, 2023)



cytb线粒体基因标记的全球哺乳动物遗传多样性预测
(Theodoridis et al., *Nature Communications*, 2020)

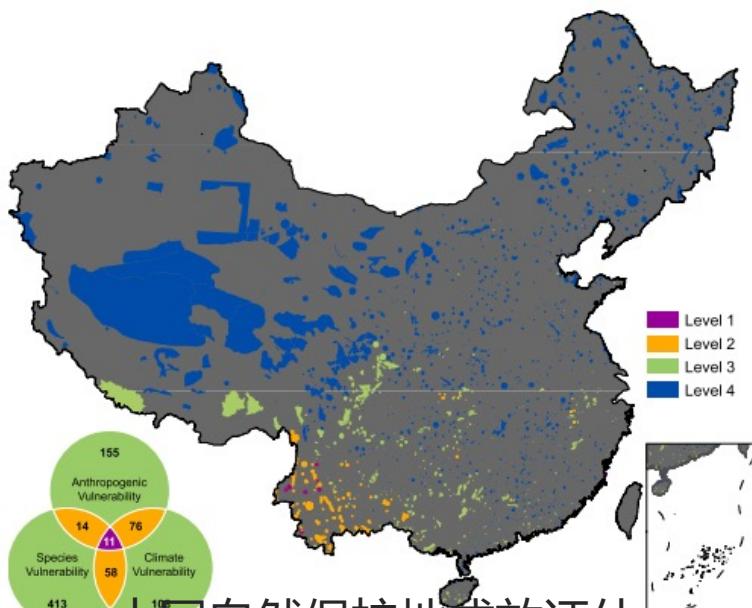
研究现状

3 保护成效评估

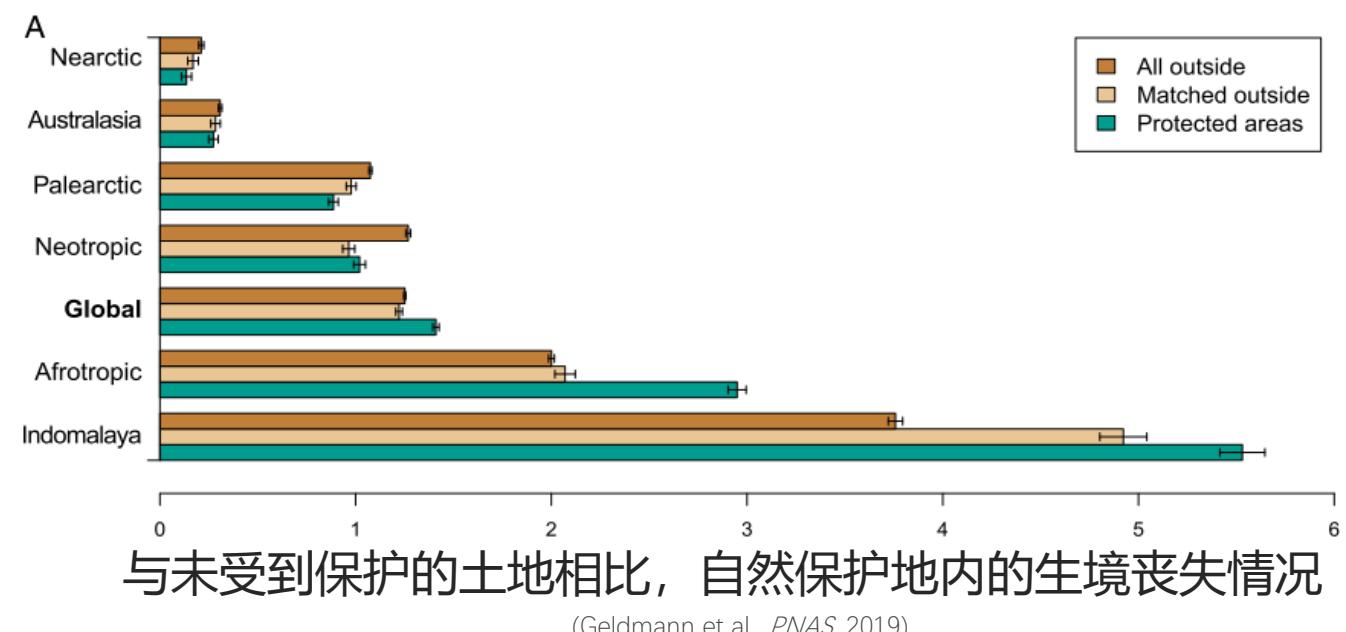
进展：揭示了保护区在缓解栖息地丧失方面的有效性和全球不同地区保护成效的显著差异

局限：

- 气候变化影响复杂且难以量化，保护成效评估的**全面性不足**
- 人类活动干扰难以分离，保护成效评估**准确性不足**
- 生物多样性动态变化难以捕捉，保护成效评估**时效性不足**



(Shrestha et al., *Nature Communications*, 2021)



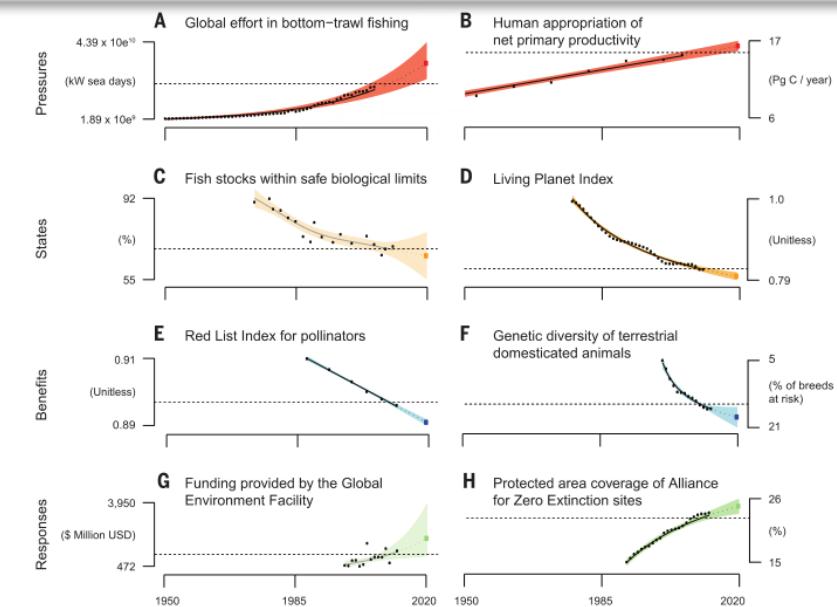
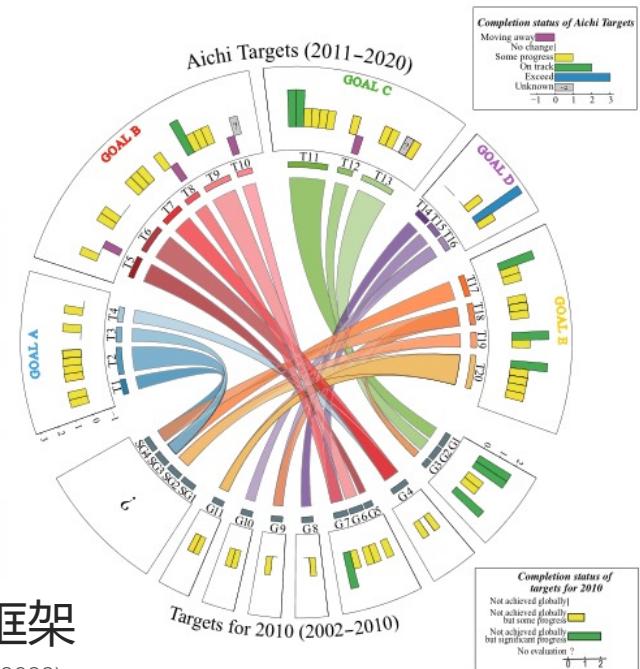
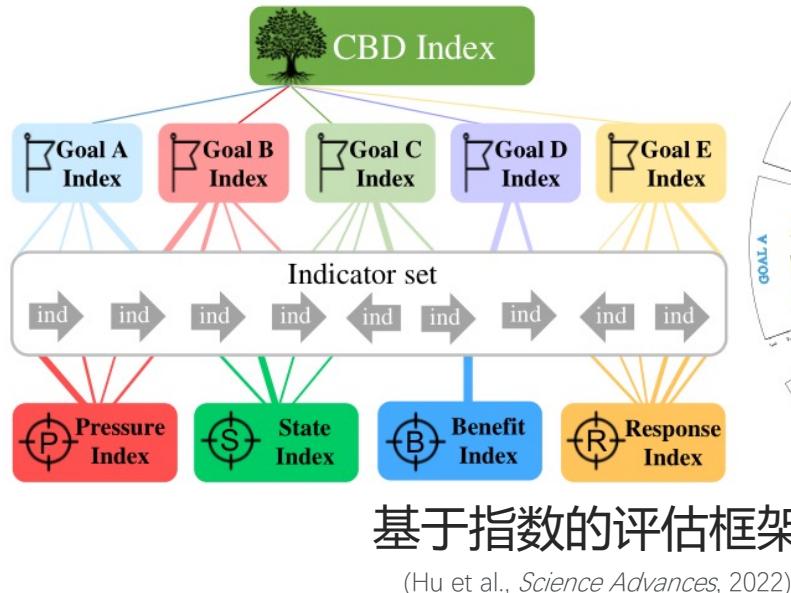
研究现状

4 全球生物多样性目标进展盘点

进展：建立基于指标的评估框架和适应性统计框架，初步揭示了在各地区和领域的不同达成情况

局限：

- 未能兼顾全面性和简洁性，指标过于复杂影响可操作性，指标过于简单影响全面性
- 因各层次目标和优先级不同，难以在全球、区域和国家多尺度上实现差异化评估
- 不同地区数据获取能力差异影响跨区域评估准确性



适应性统计框架

(Tittensor et al., *Science Advances*, 2014)

研究现状

5 AI助力生物多样性保护

从物种识别到栖息地监测，从威胁检测到决策支持，人工智能在生物多样性保护中的应用代表了一场范式转变，为解决复杂的生态挑战提供了新的视角和工具。

但面临数据质量不齐、AI偏见与幻觉等问题，未来应学科交叉、整合技术，突破数据壁垒，共建生物多样性基础大模型。

物种识别与分类



(Tomer et al., *IET Computer Vision*, 2024)

威胁预警与防范



(Kwame Nti et al., *Sustainable Futures*, 2022)

生态预测与决策支持



(Chen et al., *Ecology and Evolution*, 2021)

栖息地监测与分析



(Chen et al., *Advances in Climate Change Research*, 2022)

种群跟踪与统计

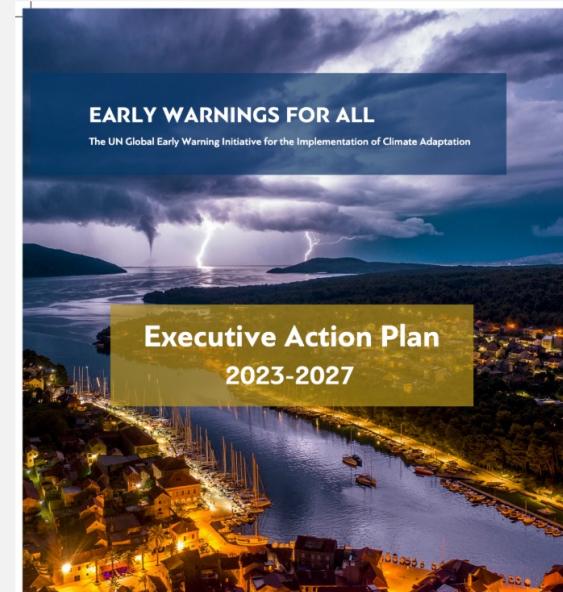


(Liu et al., *Computers and Electronics in Agriculture*, 2024)

研究现状

联合国全民早期预警行动计划

Early Warnings for All: Executive Action Plan 2023-2027



“The facts are clear. Early warnings save lives and deliver vast financial benefits. I urge all governments, financial institutions and civil society to support this effort.” – UN Secretary-General António Guterres

The Action Plan calls for investments of US\$ 3.1 billion over five years – just 50 cents per person per year – to strengthen disaster risk knowledge and management, observation and forecasting, dissemination and communication of warnings, and preparedness and response capabilities.

It leverages existing pooled funding mechanisms, such as the Climate Risk and Early Warning Systems initiative and the Systematic Observations Financing Facility, as well as global multilateral funds including the Green Climate Fund and the development banks.

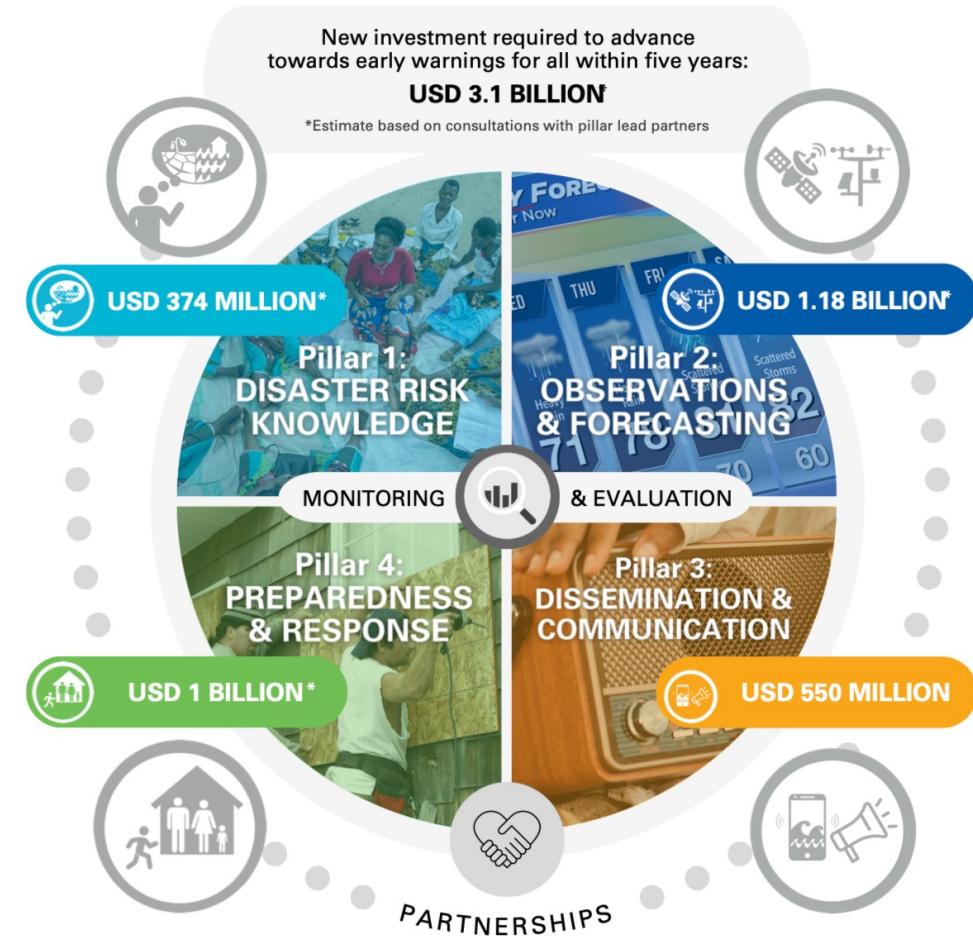


Figure 1: Budget overview for the four Pillars of the Early Warnings for All Initiative

研究现状

全球早期预警系统建设核心要素

定义：识别**潜在危机**（饥荒、洪灾、疾病等）、发出警报的信息系统，核心价值为提前识险、快速响应、减轻灾害冲击

- **风险知识**：系统性数据收集与风险评估
- **监测预警服务**：技术监测与事件预测
- **信息传播沟通**：预警信息及时传达
- **响应能力**：社区 / 公众理解与应对能力

领域	核心系统 / 倡议	关键功能
农业	GIEWS (FAO) 、 FEWS NET (USDA) 、 CropWatch (CAS) 等	作物监测、产量预估、粮食安全风险预警
城市	MCR2030 倡议、 C40 城市气候联盟	城市韧性建设、热浪 / 暴雨等气候风险应对
灾害	GMAS (WMO) 、 GDACS (UN / 欧盟)	多灾种监测、跨区域协同； 108 国已部署 (2024)
健康	EWARS (WHO) 、 GPHIN、 GOARN	传染病监测、跨国应急响应、公共卫生安全保障
生物多样性	借鉴多领域经验构建一体化框架	监测 - 评估 - 响应协同，支撑《昆蒙框架》目标

研究现状

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Earth System
Dynamics
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Review

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Plants People Planet PPP
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OPINION

Integrating machine learning, remote sensing and citizen science to create an early warning system for biodiversity

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Societal Impact Statement

The development and implementation of conservation actions that address urgent threats to biodiversity across the globe require large amounts of historic and current data. Machine learning approaches offer the tools to process, analyse and interpret large data sets, giving insights into trends and guiding evidence-based allocation of limited resources to maximise positive biodiversity outcomes. Here we describe how data acquired from remote sensing, citizen science and other monitoring approaches could feed in near-real time into an early warning system for biodiversity that integrates automated red-listing of species with the identification of priority areas for conservation.

Summary

Application of machine learning approaches is aiding biodiversity conservation and research at a time of rapid global change. Two emerging topics and their data requirements are presented. First, to identify areas of priority protection for preventing biodiversity loss, reinforcement learning is used by training models that take into account human disturbance and climate change under recurrent monitoring schemes. Second, neural networks are used to approximate classification of species into Red List categories of the International Union for Conservation of Nature, offering the possibility of real-time re-classification after events such as widespread fires and deforestation. We discuss how the identification of areas and species most at risk could be integrated into an 'early warning system' based on climatic monitoring, remotely sensed land-use changes and near-real time biological and threat data from citizen science initiatives. Such system would help guide actions to prevent biodiversity loss at the speed required for effective conservation.

KEY WORDS

artificial intelligence, citizen science, climate change, conservation biology, machine learning, modelling, neural network, warning system

Tipping point detection and early warnings in climate, ecological, and human systems

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融合多源卫星遥感与近地表相机的物候监测优化

近地面遥感是指利用**地基**传感器或成像设备来监测并研究局部**植被物候和生态系统过程**

主要优势

- 高时空分辨率（从小时级到逐日级别的观测结果），捕捉快速的物候变化
- 观测数据可与单个被监测植被构建直接联系
- 可以验证和丰富大尺度卫星监测结果



$$GCC = \frac{2 \times Green}{Red + Green}$$

绿色色度坐标

地表监测应用

- 可与卫星图像相结合，进行高分辨率、高频率的土地覆盖动态制图
- 监测植被和作物状况，验证卫星遥感监测产品，了解多尺度生态系统过程

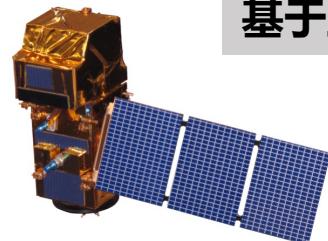
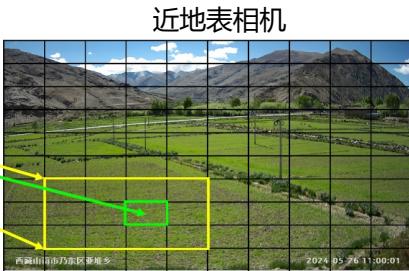
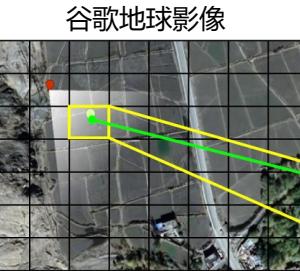
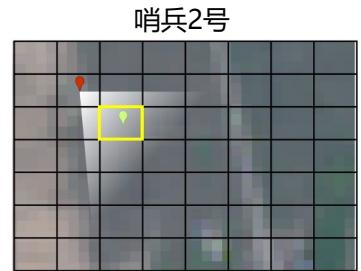


提供验证数据：

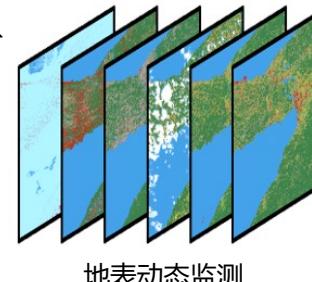
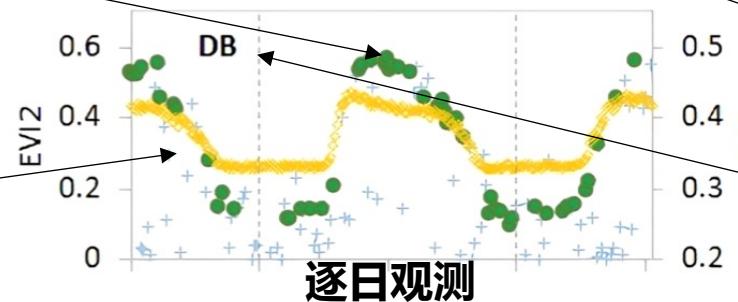
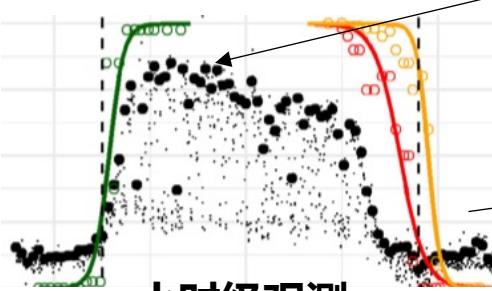
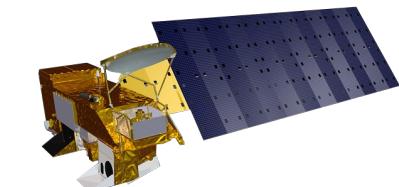
- 植被物候
- 人类活动
- 积雪变化
- 极端天气
-

融合多源卫星遥感与近地表相机的物候监测优化

近地表相机与卫星数据的空间匹配及融合



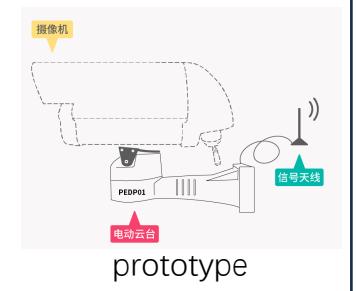
基于卫星的高质量无云逐日观测受限 (~2%)



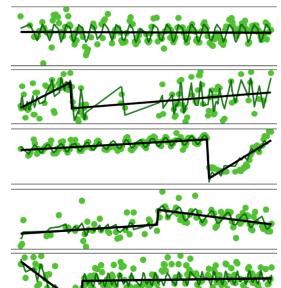
地表动态监测

新型近地表相机设计

- 网络状况
- 天气状况
- 能源效率
- 易于安装
- 图像/视频/音频
- 价格低廉

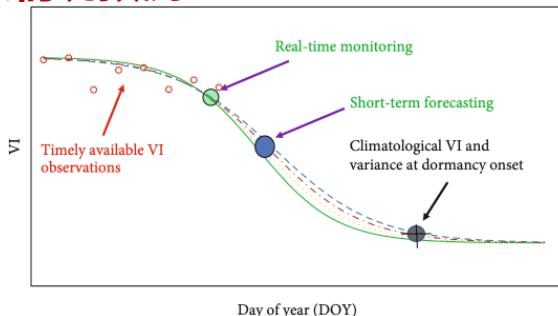
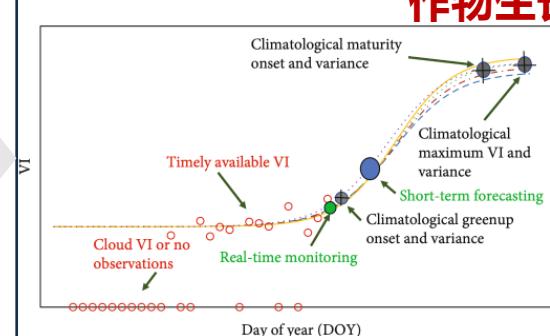
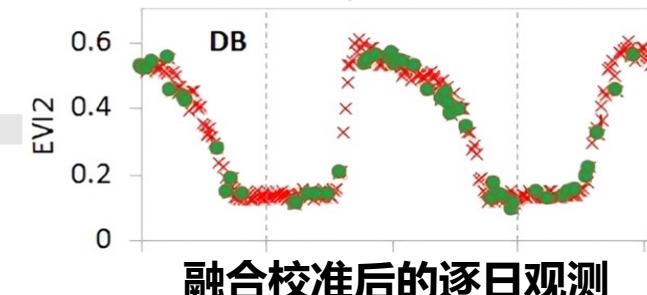


精准且快速的地表动态变化监测



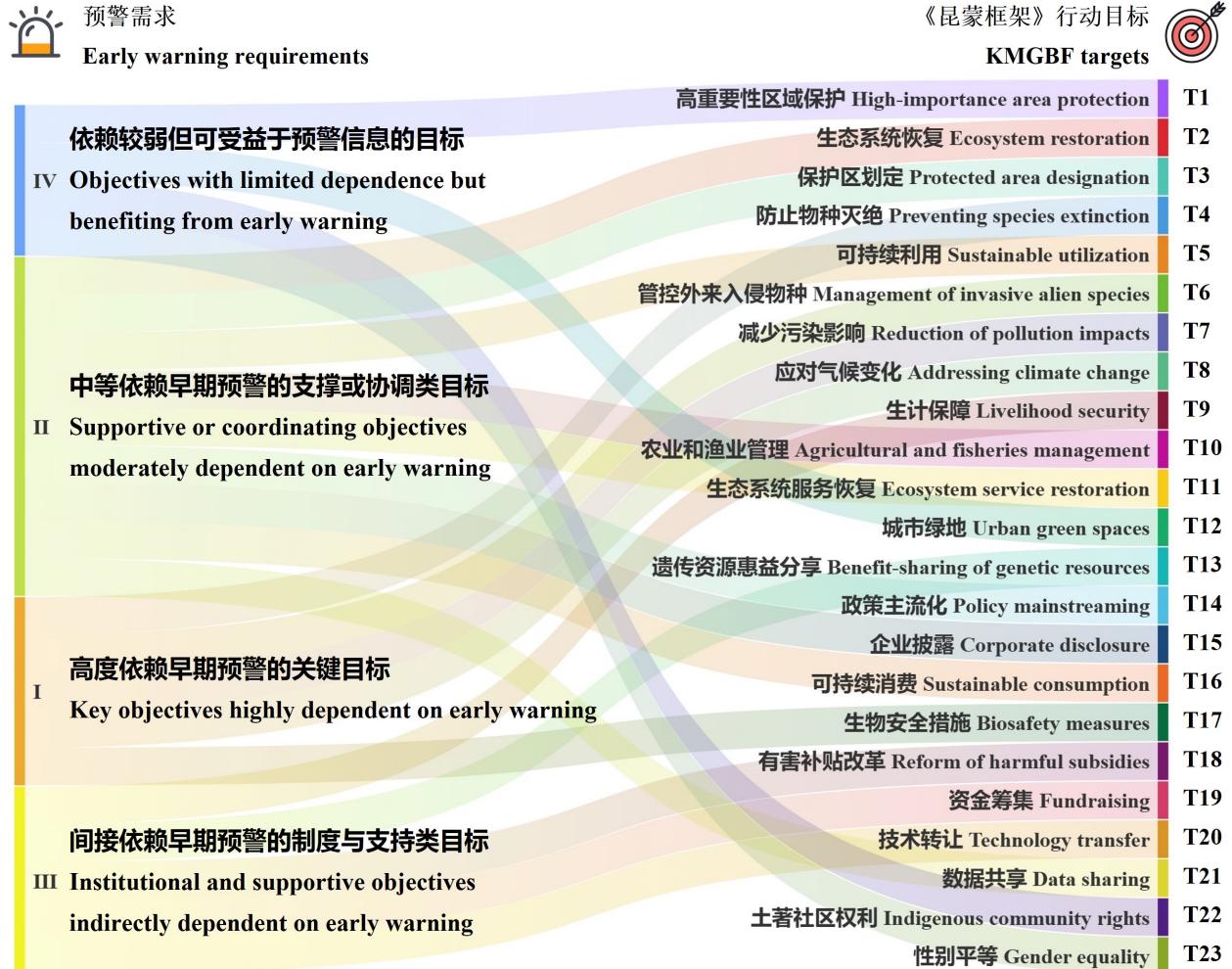
变化程度
(细微变化, 巨变)

变化持续时段
(突变, 缓变)

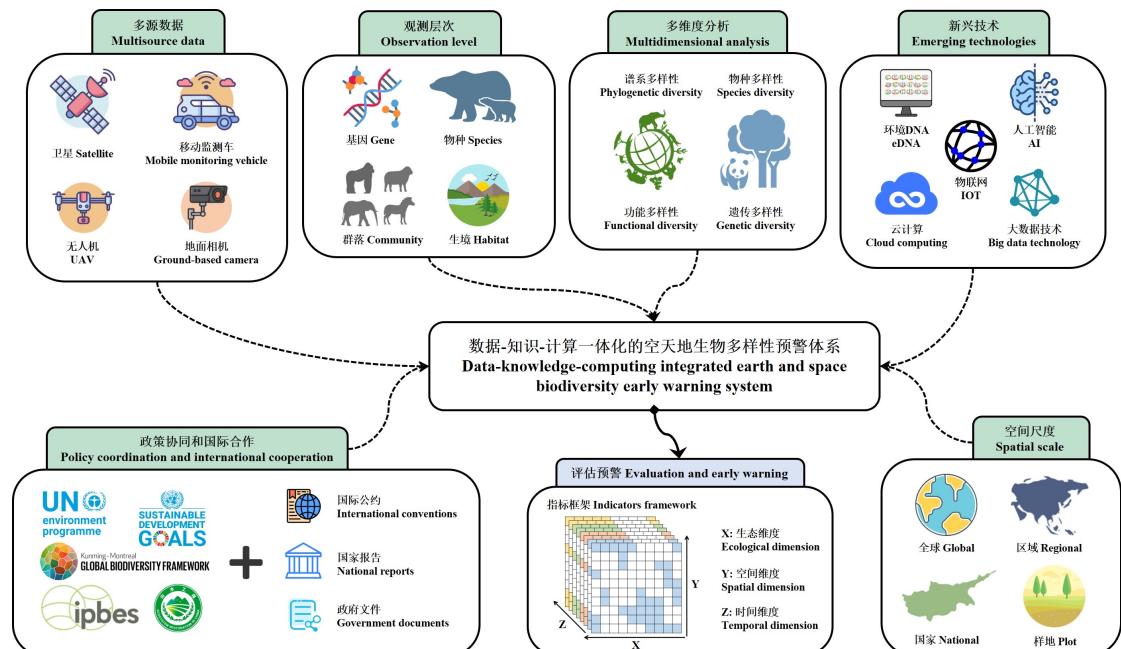


生物多样性变化快速评估与早期预警

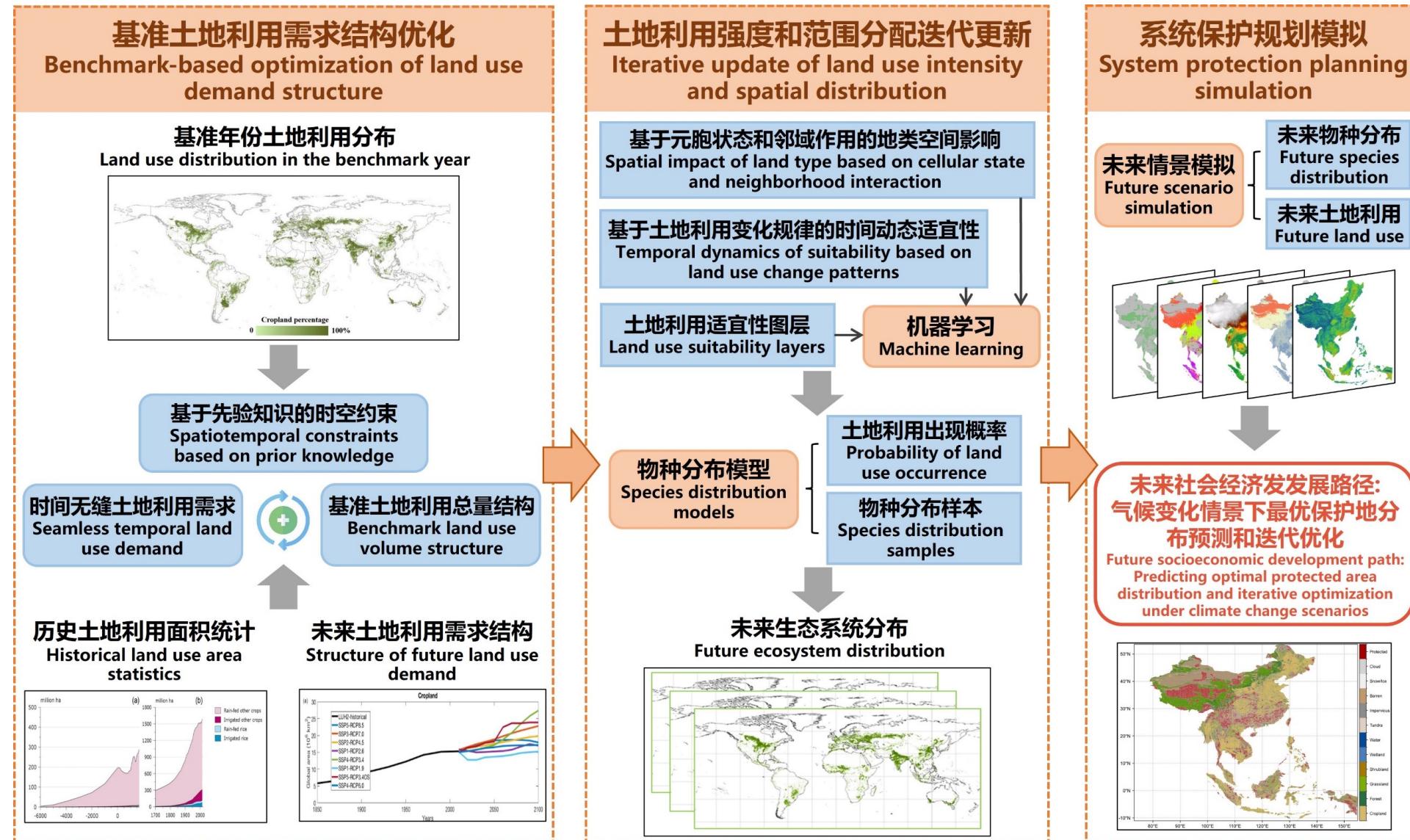
面向《昆蒙框架》目标的生物多样性早期预警



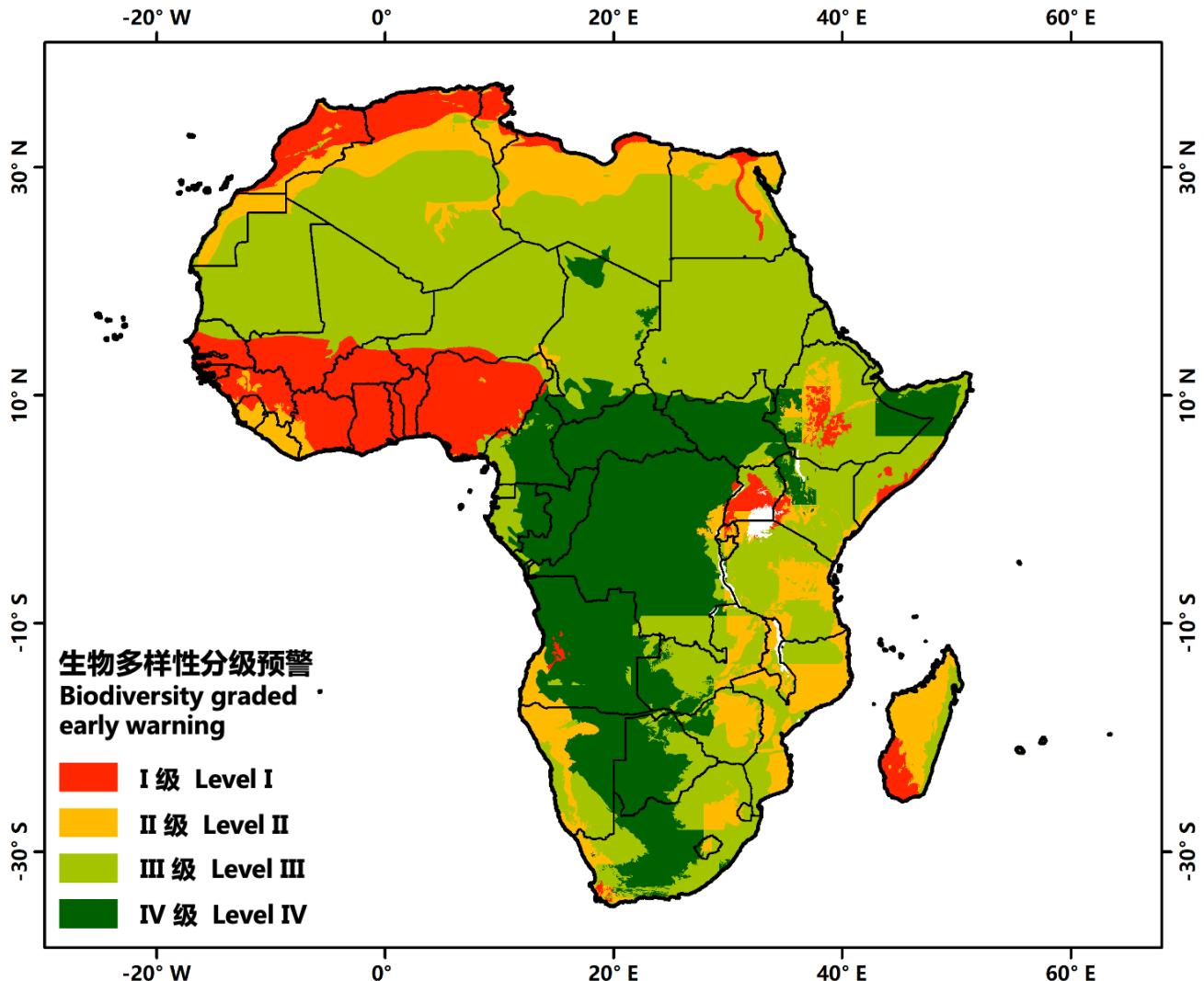
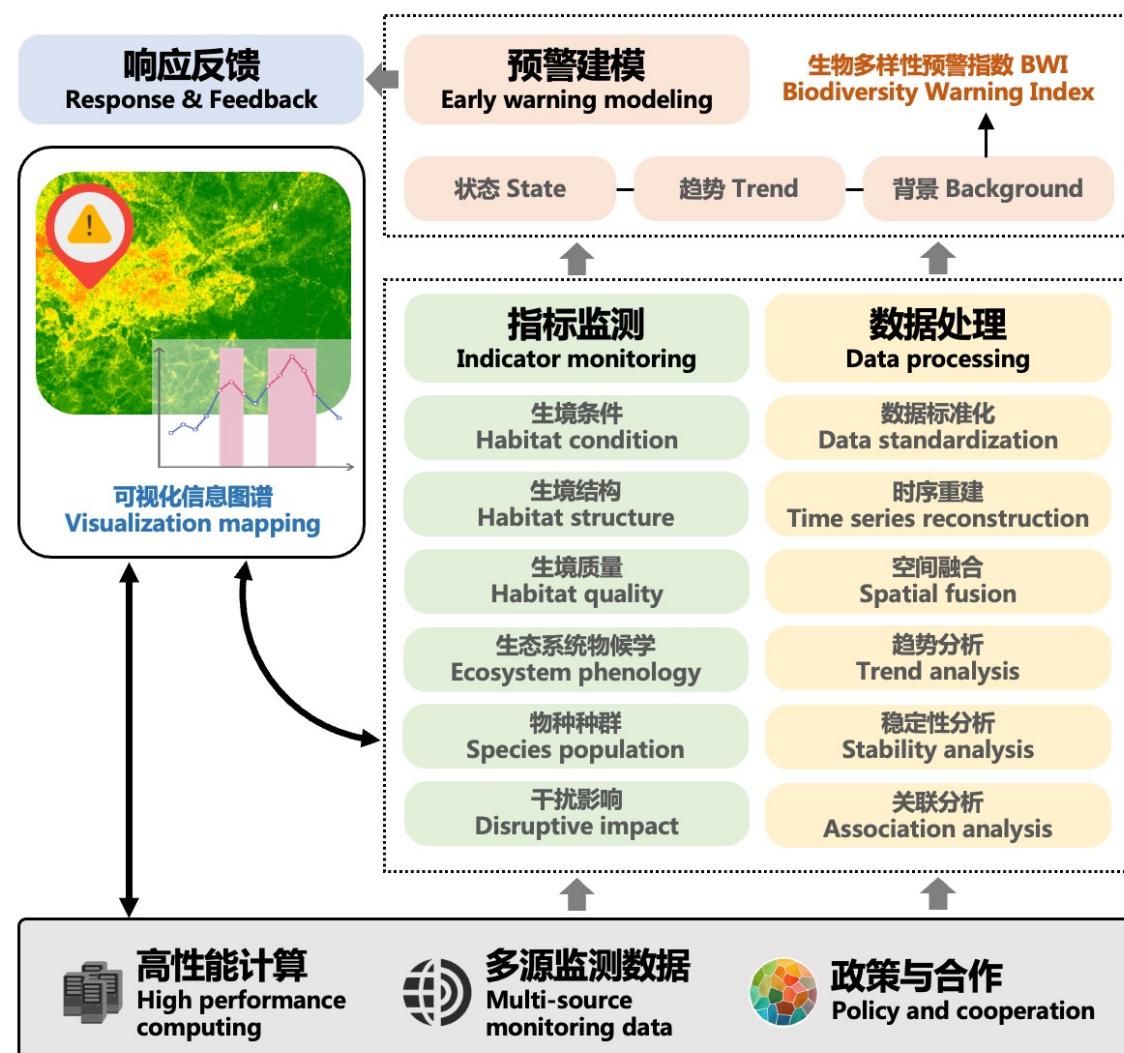
面向《昆明-蒙特利尔全球生物多样性框架》的全球生物多样性预警体系



生物多样性变化快速评估与早期预警



生物多样性变化快速评估与早期预警



致谢：武慧、杜贞容、顾劬同、赵强、李曦煜

感谢各位专家
敬请批评指正

俞乐

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2025-12-25



Kunming-Montreal

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