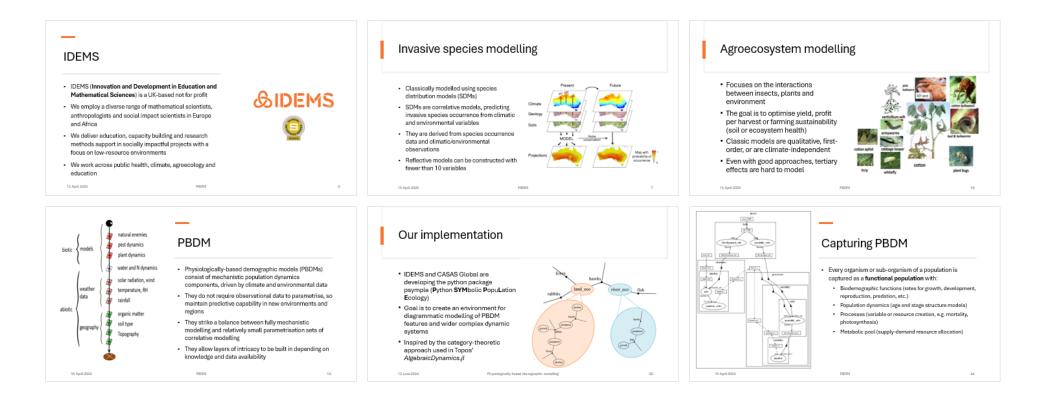
Physiologically-based demographic modelling

A compositional modelling approach

George Simmons, IDEMS

Today's structure



IDEMS

- IDEMS (Innovation and Development in Education and Mathematical Sciences) is a UK-based not for profit
- We employ a diverse range of mathematical scientists, anthropologists and social impact scientists in Europe and Africa
- We deliver education, capacity building and research methods support in socially impactful projects with a focus on low-resource environments
- We work across public health, climate, agroecology and education

GIDEMS



CASAS Global





- CASAS (**Center for Analysis of Sustainable Agroecological Systems**) is a nonprofit organisation of collaborating scientists founded by Prof. Andrew Gutierrez (UC Berkeley)
- Their mission is dedicated to analysing issues in diverse crops, rangelands and medical and veterinary vectors to benefit populations and governments in developing countries worldwide

Collaboration aims

- Produce an open-source codebase implementing a general, reusable and modular of CASAS Global's physiologically-based demographic modelling (PBDM) framework
- Build an accessible platform of tools for general users to build, parametrise and simulate PBDMs
- Allow for third-party solutions and expertise to integrate with the platform
- Build scalability through internship programmes, education, training, and outreach schemes
- Maximise the impact of these modelling ideas by building the right solutions, interfaces and integrations for stakeholders

What we're currently doing

Software

- A python-based, open-source compositional modelling package implementing population dynamics and functional responses
- An open-source, general and reusable PBDM implementation

Scaling & outreach

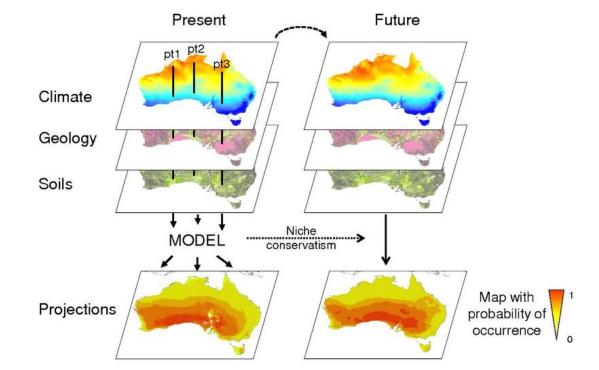
- An internship programme in Kenya to teach PBDM, driving local knowledge and creating collaborations
- Support for locally-relevant Master's projects around insect and crop modelling

Accessibility & impact

- Building for interoperability, i.e. allowing different crop modelling systems or climate data sources to integrate with the platform
- An emphasis on building the right solutions for different users (academic, government, local/regional policy, industry)

Invasive species modelling

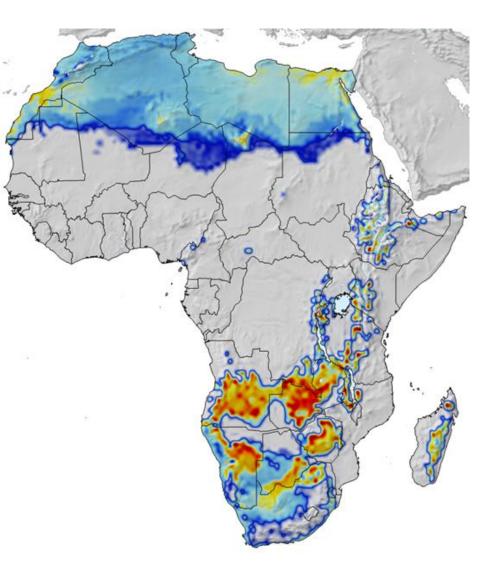
- Classically modelled using species distribution models (SDMs)
- SDMs are correlative models, predicting invasive species occurrence from climatic and environmental variables
- They are derived from species occurrence data and climatic/environmental observations
- Reflective models can be constructed with fewer than 10 variables

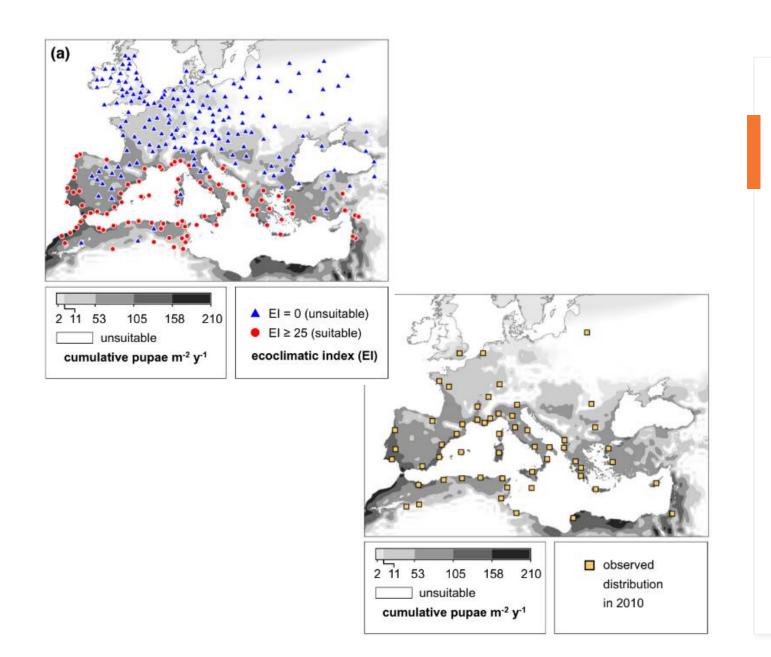


Regional modelling

Possible questions:

- What potential invasive species have to thrive in different regions
- Predicting disease vectors, agricultural impacts and economic costs
- How to use regional policy to have the biggest impact potential
- How these factors change with climate change



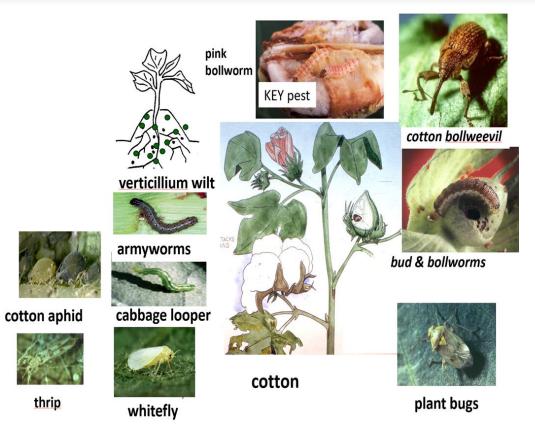


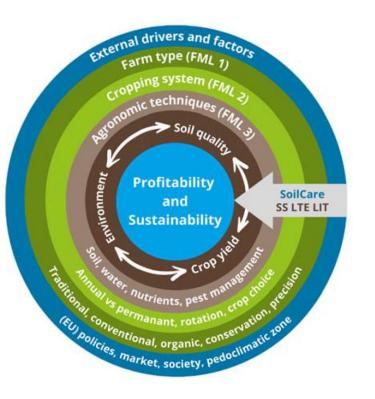
Case study: tomato pinworm (part 1)

- Correlative models predicted tomato pinworm distribution in Europe in 2010 to be limited to Mediterranean boundary
- Based on occurrence data from South America
- Observations in 2010 were recorded around the Black Sea, Baltics, France and UK

Agroecosystem modelling

- Focuses on the interactions between insects, plants and environment
- The goal is to optimise yield, profit per harvest or farming sustainability (soil or ecosystem health)
- Classic models are qualitative, firstorder, or are climate-independent
- Even with good approaches, tertiary effects are hard to model





Agronomic modelling

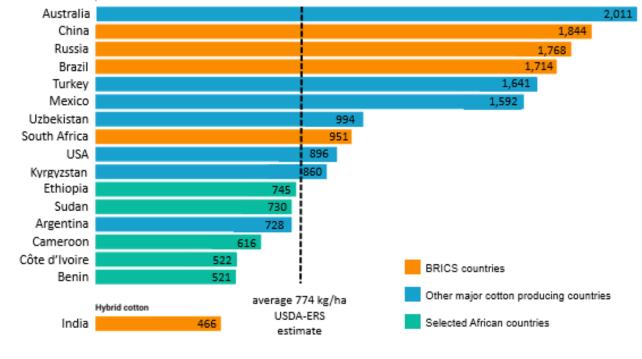
Possible questions:

- Optimal planting methods and farming techniques
- How to control pests (biocontrol, traps, pesticides, crop hygiene)
- Understand soil health or wider environmental impacts
- The impacts of weather and climate change
- The impact of government regulations or subsidies

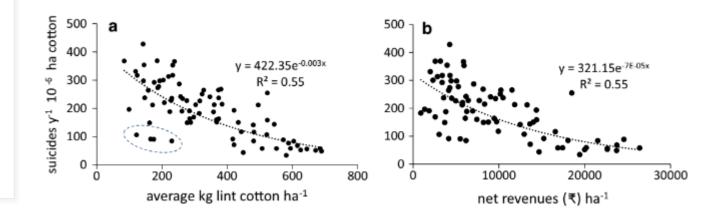
Case study: cotton in India (part 1)

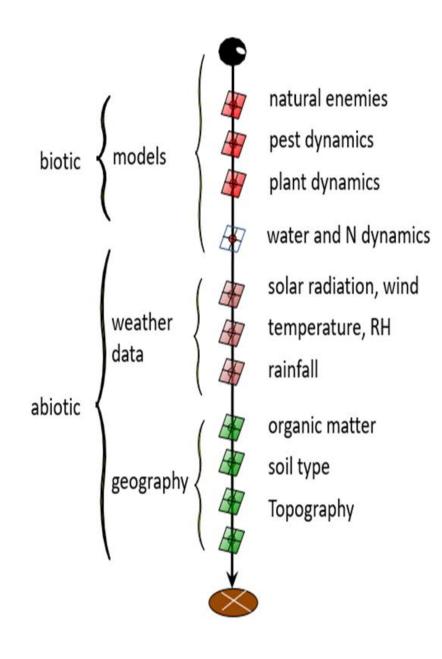
- Hybrid genetically-modified cotton was introduced in India, producing a toxin to control pink bollworm
- After initially good results, yields stagnated and declined
- More than 100,000 suicides attributed to low cotton yields and failures

Average national seed cotton yields (kg/ha) 2020-2021



2021 data - International Cotton Advisory Committee, Washington DC



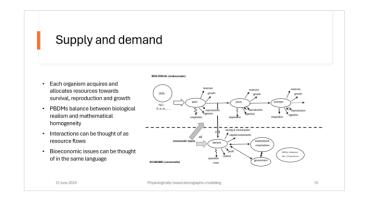


PBDM

- Physiologically-based demographic models (PBDMs) consist of mechanistic population dynamics components, driven by climate and environmental data
- They do not require observational data to parametrise, so maintain predictive capability in new environments and regions
- They strike a balance between fully mechanistic modelling and relatively small parametrisation sets of correlative modelling
- They allow layers of intricacy to be built in depending on knowledge and data availability

Three key features of PBDM

Demand-driven evolution



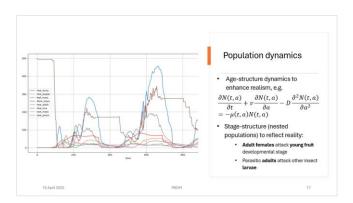
Weather and climate dependency

 $\phi_{\mathcal{B}} \leq 1$:

15 April 2025

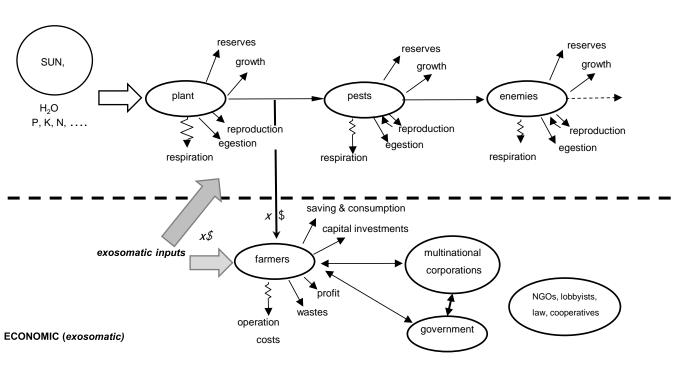
Environmental response · The effects of climate (temperature, humidity, day length) or other factors (age, density) are captured by functions parametrised from field of laboratory data Responses to other variables are captured by scalar multipliers 0 ≤ $rate(t) = \prod \phi_{g}(t) \times rate_{opt}(t)$ PBDM

Age and stagestructured dynamics



Supply and demand

- Each organism acquires and allocates resources towards survival, reproduction and growth
- PBDMs balance between biological realism and mathematical homogeneity
- Interactions can be thought of as resource flows
- Bioeconomic issues can be thought of in the same language

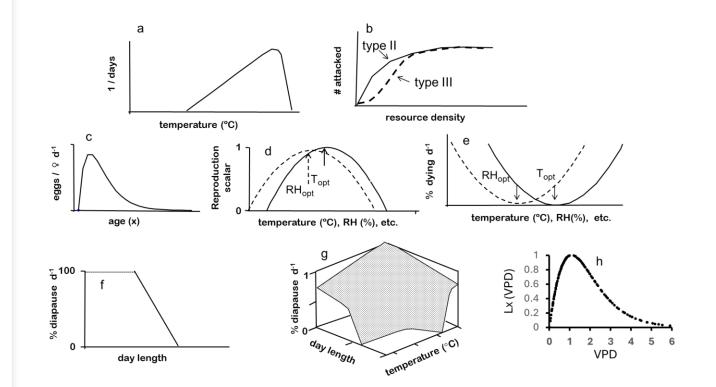


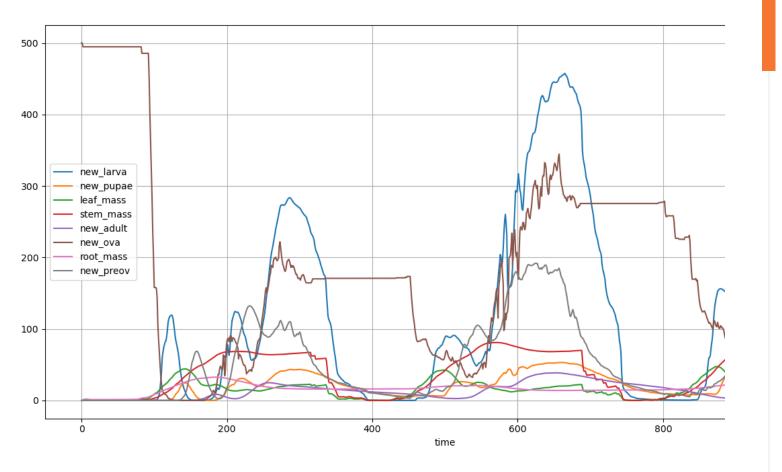
BIOLOGICAL (endosomatic)

Environmental response

- The effects of climate (temperature, humidity, day length) or other factors (age, density) are captured by functions parametrised from field or laboratory data
- Responses to other variables are captured by scalar multipliers $0 \le \phi_E \le 1$:

$$rate(t) = \prod_{E} \phi_{E}(t) \times rate_{opt}(t)$$



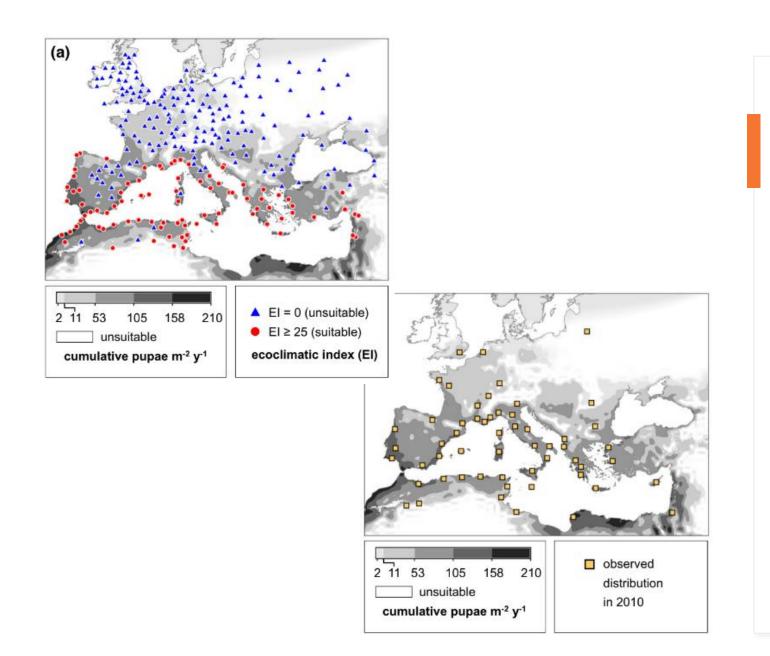


Population dynamics

• Age-structure dynamics to enhance realism, e.g.

$$\frac{\partial N(t,a)}{\partial t} + v \frac{\partial N(t,a)}{\partial a} - D \frac{\partial^2 N(t,a)}{\partial a^2} \\ = -\mu(t,a)N(t,a)$$

- Stage-structure (nested populations) to reflect reality:
 - Adult females attack young fruit developmental stage
 - Parasitic adults attack other insect
 larvae



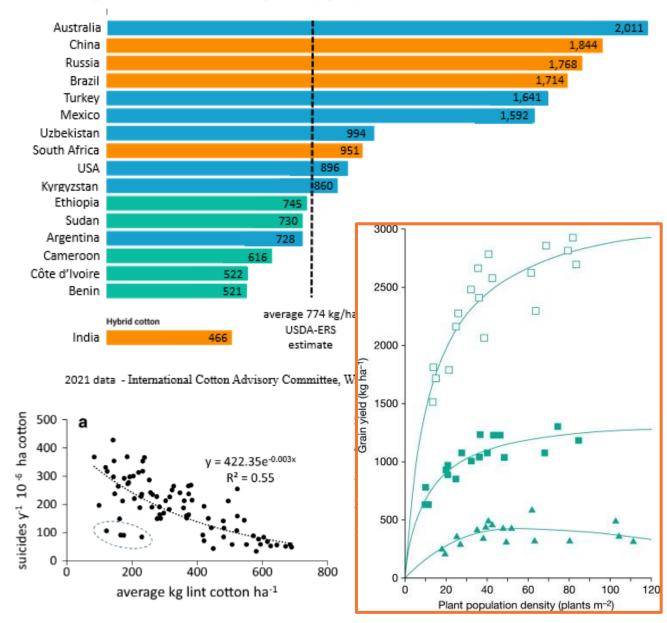
Case study: tomato pinworm (part 2)

- PBDM analysis (greyscale) predicts the observational data much more closely than the correlative data based on South American observations
- Since PBDM does not require the occurrence data to parametrise, they would have predicted the 2010 outbreak of *T. absoluta* in the EU

Case study: cotton in India (part 2)

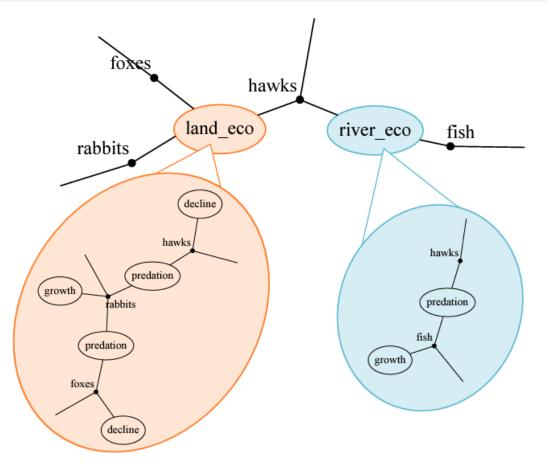
- PBDM analysis indicates that lower planting density caused by higher seed cost leads to:
 - lower yields as the plant prioritises vegetative growth to fill its space
 - An enlongated growing season, leading to susceptibility to other pests
- Increased resistance over time
- Farmers locked into a cycle

Average national seed cotton yields (kg/ha) 2020-2021



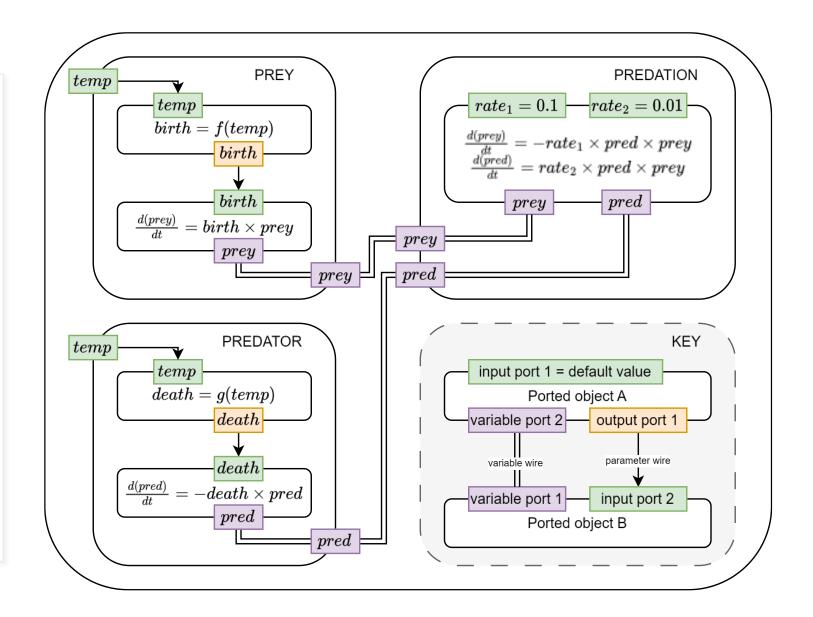
Our implementation

- IDEMS and CASAS Global are developing the python package psymple (Python SYMbolic PopuLation Ecology)
- Goal is to create an environment for diagrammatic modelling of PBDM features and wider complex dynamic systems
- Inspired by the category-theoretic approach used in Topos' *AlgebraicDynamics.jl*



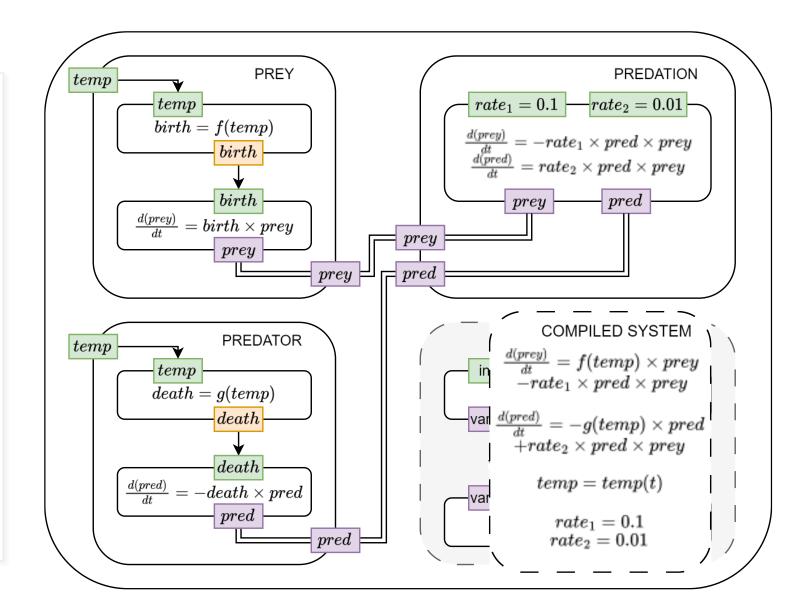
Example: temperaturedependency

- Diagrammatic representation of predatorprey
- In this form, it's easy to add temperature-dependent birth and mortality functions



Example: temperaturedependency

- Compilation collects equations connected by variable wires and aggregates them
- Functions connected by parameter wires undergo partial substitution



12 June 2024

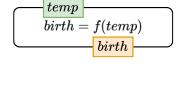
Ported objects as code

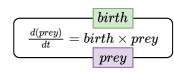
This system represents the differential equation

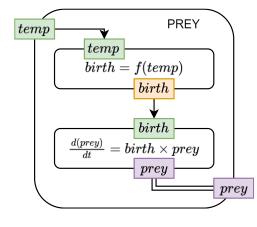
$$\frac{d(prey)}{dt} = f(temp) * prey$$

where

$$f(temp) = 1 - temp^2$$







Birth

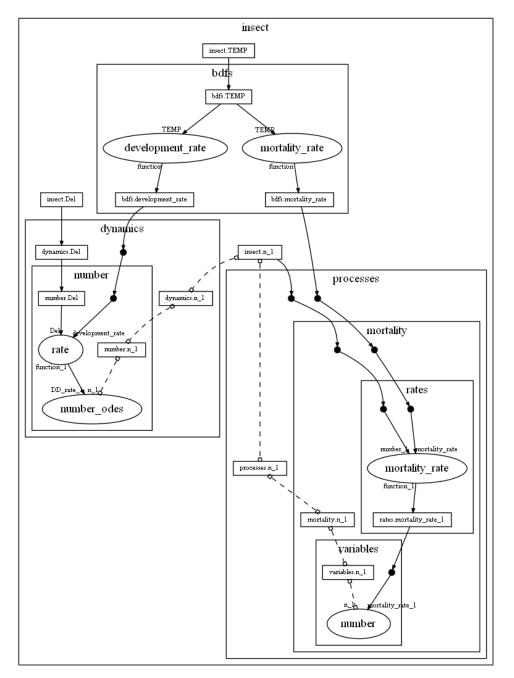
birth = FunctionalPortedObject(
 name="birth_rate",
 assignments=("birth", "1 - temp**2")
)

Prey dynamics

```
prey_dyn = VariablePortedObject(
    name="dyn",
    assignments=[("prey", "birth*prey")],
)
```

Prey functional population

prey = CompositePortedObject(
 name="prey",
 children=[birth, prey_dyn],
 variable_ports=["prey"],
 input_ports=["temp"],
 directed_wires=[
 ("temp", ["birth_rate.temp"]),
 ("birth_rate.birth", ["dyn.birth"]),
],
 variable_wires=[(["dyn.prey"], "prey")],

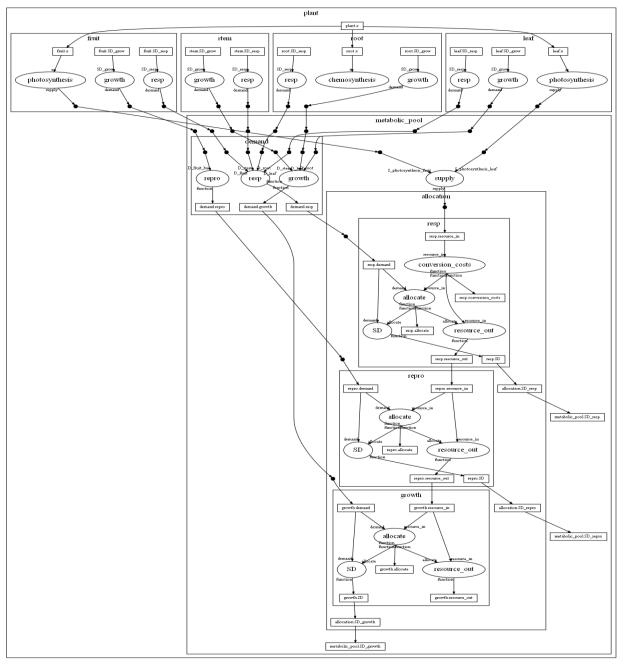


Capturing PBDM

- Every organism or sub-organism of a population is captured as a **functional population** with:
 - Biodemographic functions (rates for growth, development, reproduction, predation, etc.)
 - Population dynamics (age and stage structure models)
 - Processes (variable or resource creation, e.g. mortality, photosynthesis)
 - Metabolic pool (supply-demand resource allocation)

Metabolic pool

This object calculates a set of scalars, there are no variables involved!



```
##### Insect functional population #####
```

```
F = FunctionalPopulation(
    name="insect",
    age_structure = {
        "k": 5,
    },
    Del = 220,
    variable = "n",
    bdfs = {
        "development_rate": {
            "function": "a*(TEMP - T min)/(1 + b**(TEMP - T max))",
            "inputs": {"a": 1, "b": 2.5, "T_min": 9.5, "T_max": 35.5}
        },
        "mortality rate": {
            "function": "zerone(a*TEMP**2 + b*TEMP + c)",
            "inputs": {"a": 0.00098, "b": -0.0098, "c": 0.0001}
        }
    },
    dynamics =
        "number": {
            "type": "distributed delay",
            "rate": {
                "function": "development rate * k/Del",
                "inputs": {"development rate": "insect.bdfs.development rate"}
    },
    processes =
        "mortality": {
            "rates": {
                "mortality rate": {
                    "type": "age_structured",
                    "function": "mortality_rate * number",
                    "inputs": {"mortality_rate": "insect.bdfs.mortality_rate"},
                    "age_structured_inputs": {"number": "insect.n"}
            },
            "variables":
                "number": {
                    "type": "age_structured",
                    "function": "- mortality rate",
                    "age_structured_inputs": {"mortality_rate"}
```

Other features

- Code interface for creating components of functional populations (bdfs, dynamics, processes, etc.), with internal automation to create objects and wires
- Automated hierarchical searching for parameters (pull vs. push behaviour), e.g. Del
- Facility to implement system-wide parameters (TEMP) and utility functions (zerone)
- Relative hierarchical address-based system to pull or share information with non-immediate siblings (insect.bdfs.mortality_rate)

Thank you!

- IDEMS: <u>https://idems.international</u>
- CASAS: http://casasglobal.org
- psymple:
 - Repo: https://github.com/casasglobal-org/psymple
 - Docs: <u>https://casasglobal-org.github.io/psymple</u>