



Crop Modelling CoP Webinar:
Disease crop modeling advances
and challenges for large scale
simulation studies



From the evaluation of the environmental suitability of plant pathogens to the integration of disease and crop models: data needs, use cases and challenges

Simone Bregaglio

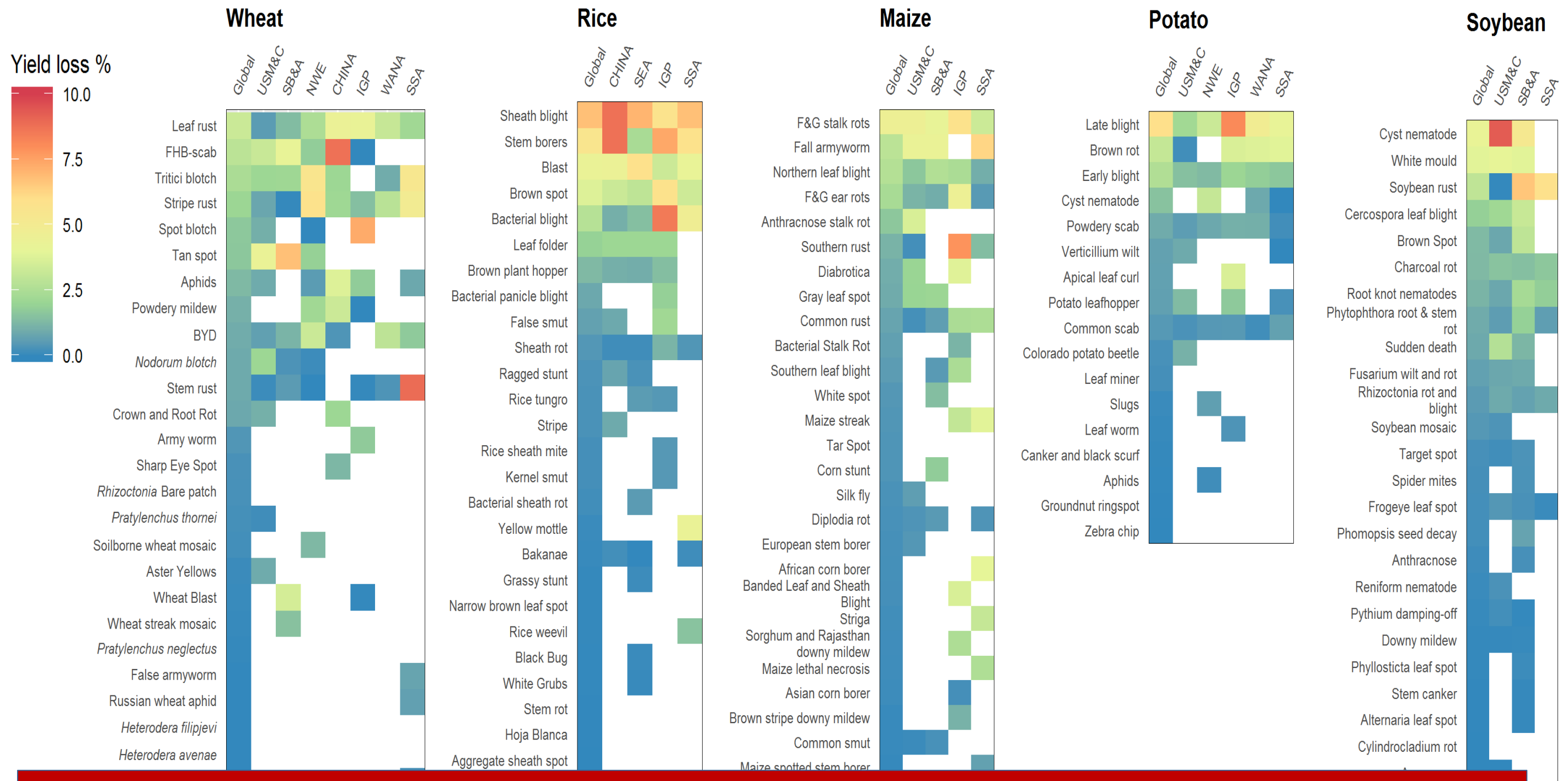
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The relevance of modelling the impact of plant pests and diseases on crop



- Plant diseases are a **major cause for yield losses** worldwide and an important, yet poorly documented, **source of uncertainty in modelling outputs**.
- Disease yield loss quantification is critical to inform **tactical to strategic decisions for disease management**, for priority setting and to support public policies.
- **Yield losses can be simulated** using agrophysiological models in which **damage mechanisms** have been incorporated.
- **Simulations** that do not account for crop pathogens and pests **are bound to lead to overestimates**, in some cases, very large ones.

Estimates of crop losses by pathogens and pests



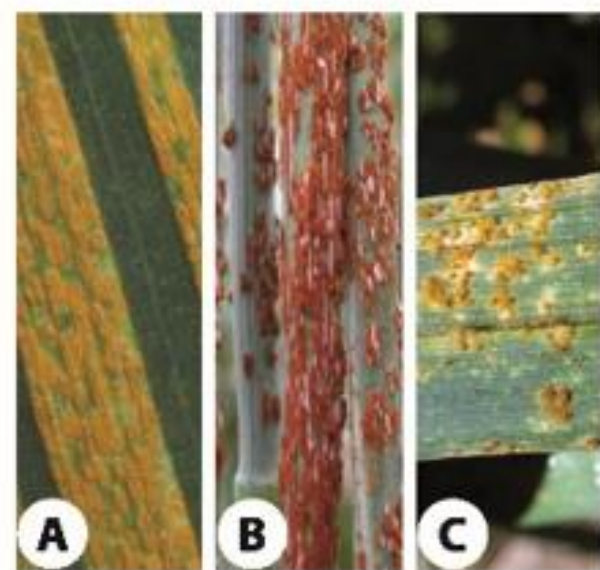
137 P&Ps were reported with a large variation in crop losses caused by specific P&Ps. The relative importance of P&Ps varied across food security hotspots

The impacts of plant disease epidemics on food security can be examined



- WHEAT RUSTS
- FUSARIUM HEAD BLIGHT
- POTATO LATE BLIGHT
- RICE BROWN SPOT

WHEAT RUSTS



Source: strip rust.nrsu.edu, theconversation.com, biog.extension.uga.edu

A) *P. striiformis*, B) *P. graminis*, C) *P. triticina*

Where: N. America, Europe, Africa, Asia

When: 20th and 21th century

Chronic, acute and emerging

Type B, C

FUSARIUM HEAD BLIGHT



Source: news.ca.uky.edu

Fusarium spp.

Where: N. America, China, Europe

When: Today

Acute

Type G

| COMPONENT | EFFECT |
|-----------------|--------------|
| Production | Large |
| Stocks | Impacted |
| Physical access | Not affected |
| Economic access | Impacted |
| Stability | Impacted |
| Nutritive value | Large |

| COMPONENT | EFFECT |
|-----------------|-------------------|
| Production | Strongly impacted |
| Stocks | Strongly impacted |
| Physical access | Impacted |
| Economic access | Impacted |
| Stability | Impacted |
| Nutritive value | Very large |

- POTATO LATE BLIGHT
- RICE BROWN SPOT
- WHEAT RUSTS

Source: freepik.com

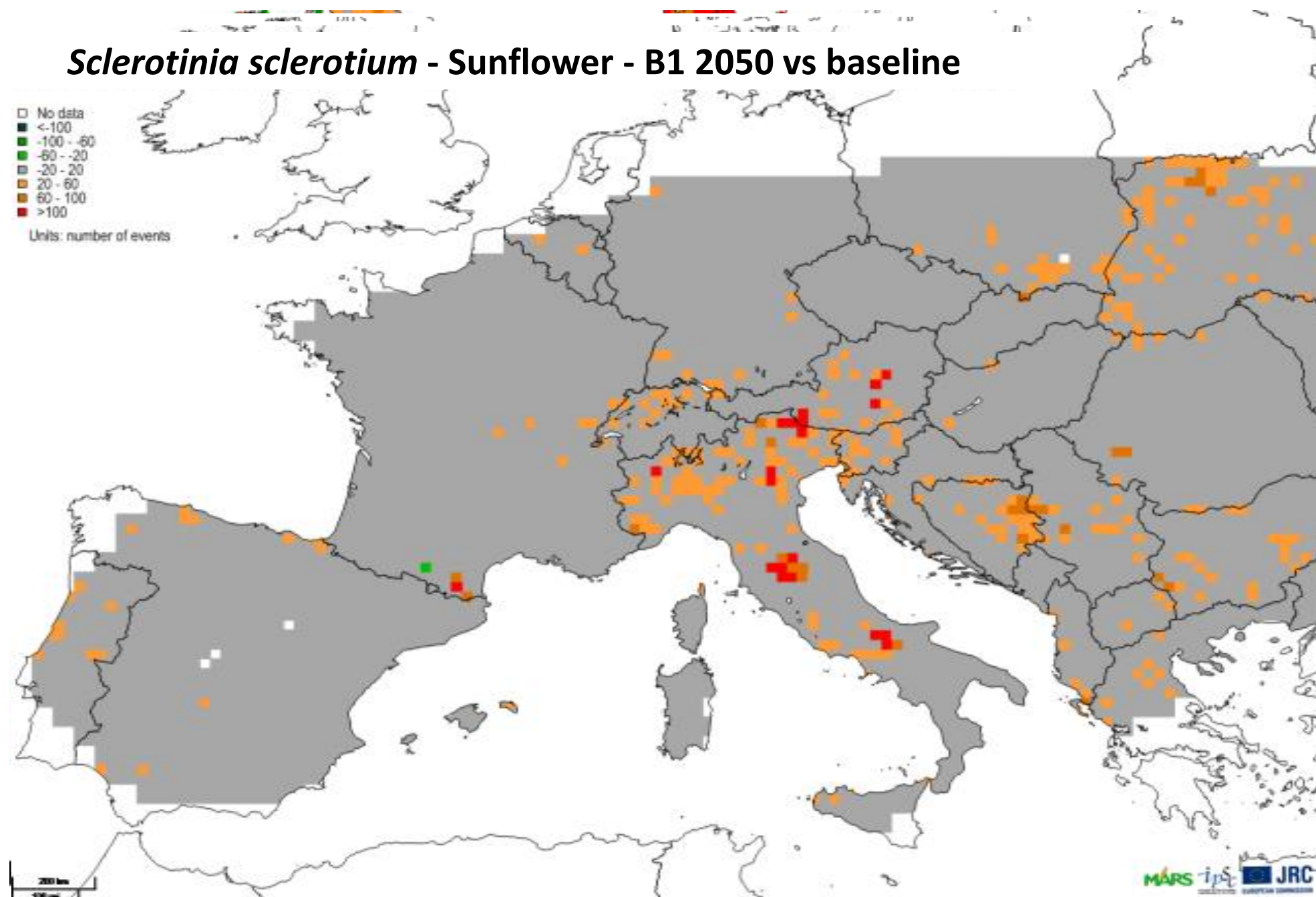
| | |
|-----------------|--------------|
| Physical access | Impacted |
| Economic access | Large |
| Stability | Not affected |
| Nutritive value | Not affected |

Savary et al., 2017,
Food Security

Climate change leads to high uncertainty in plant disease spread and intensity

- The effect of **changing atmospheric composition and climate** on individual **pathosystems** can be *positive, negative or neutral* (Garrett et al., 2006; Chakraborty et al., 2008)

Sclerotinia sclerotium - Sunflower - B1 2050 vs baseline

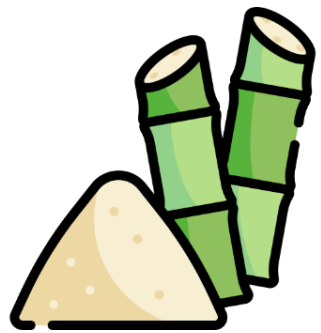


Bregaglio et al., 2012,
Agron. Sust. Dev.

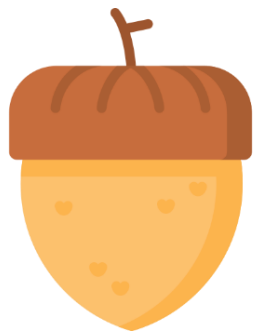
Why modelling pests and diseases impact on crop is needed?



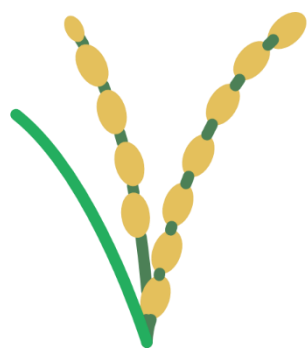
- A possible classification of plant disease models:
 - **Epidemiological models**, the target is the **dynamic** of plant disease epidemics
 - **Yield loss models**, the target is the reproduction of the **impact** of the pest/disease on crop yield.



- A model to predict the suitability of weather conditions to the orange rust disease of sugarcane.



- An informative model-based system for early forecast of hazelnut quality.

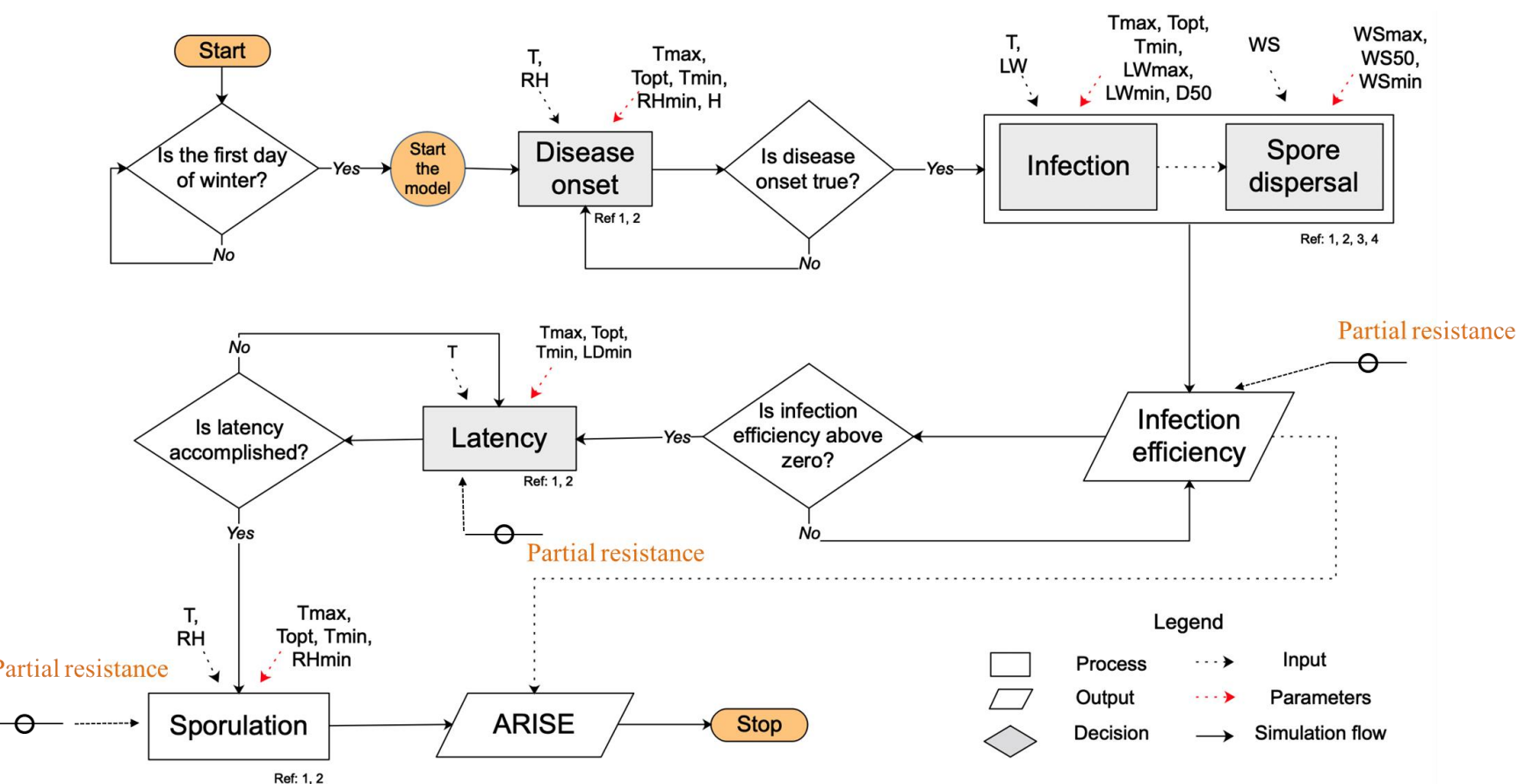


- A model to predict the impact of rice blast epidemics in current and future scenarios.

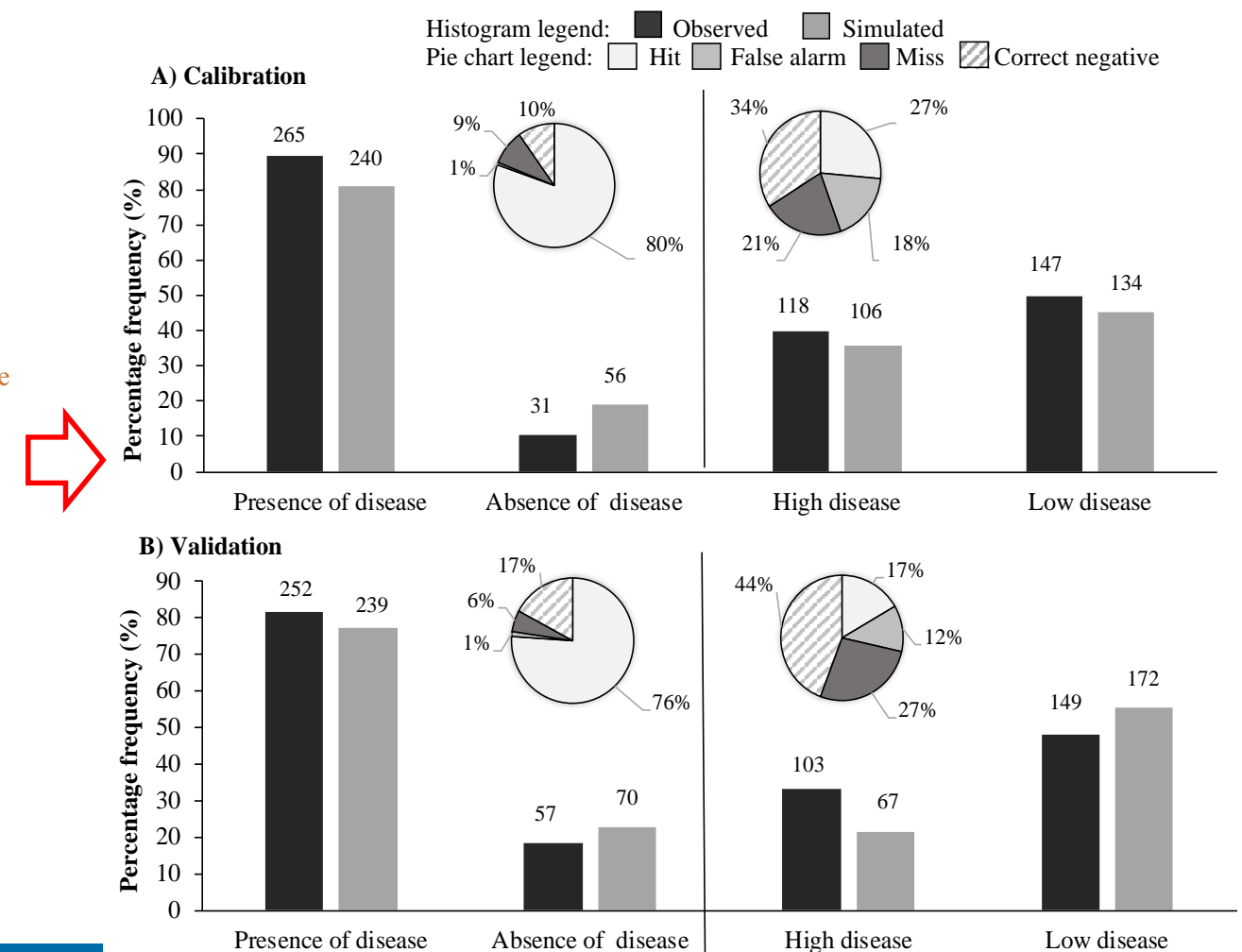
Use cases: orange rust on sugarcane

- Orange rust caused by the polycyclic fungus *Puccinia kuehnii* is an emerging disease of sugarcane.
- Yield losses in susceptible and intermediate resistant varieties can reach 40%.
- Need of forecasting systems for in-season support and scenario analyses.

Existing modelling approaches calibrated with literature data

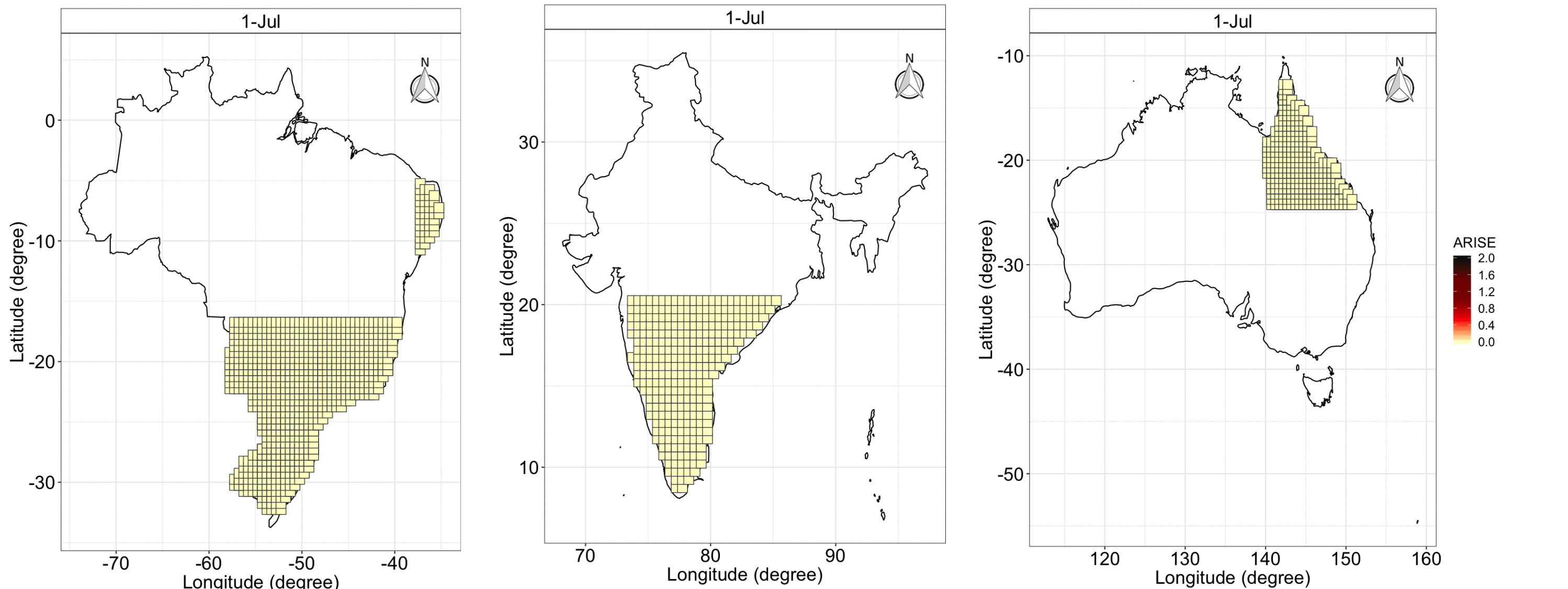


Model performances in reproducing disease severity



Use cases: orange rust on sugarcane

- A total of 605 field samplings were taken on 70 sugarcane farms in the Pradópolis region, state of São Paulo → need of upscaling according to weather database.
- After the evaluation with field data, the model was projected on main sugarcane producing country to test its behavior on large area simulations.



Use cases: rotten hazelnuts

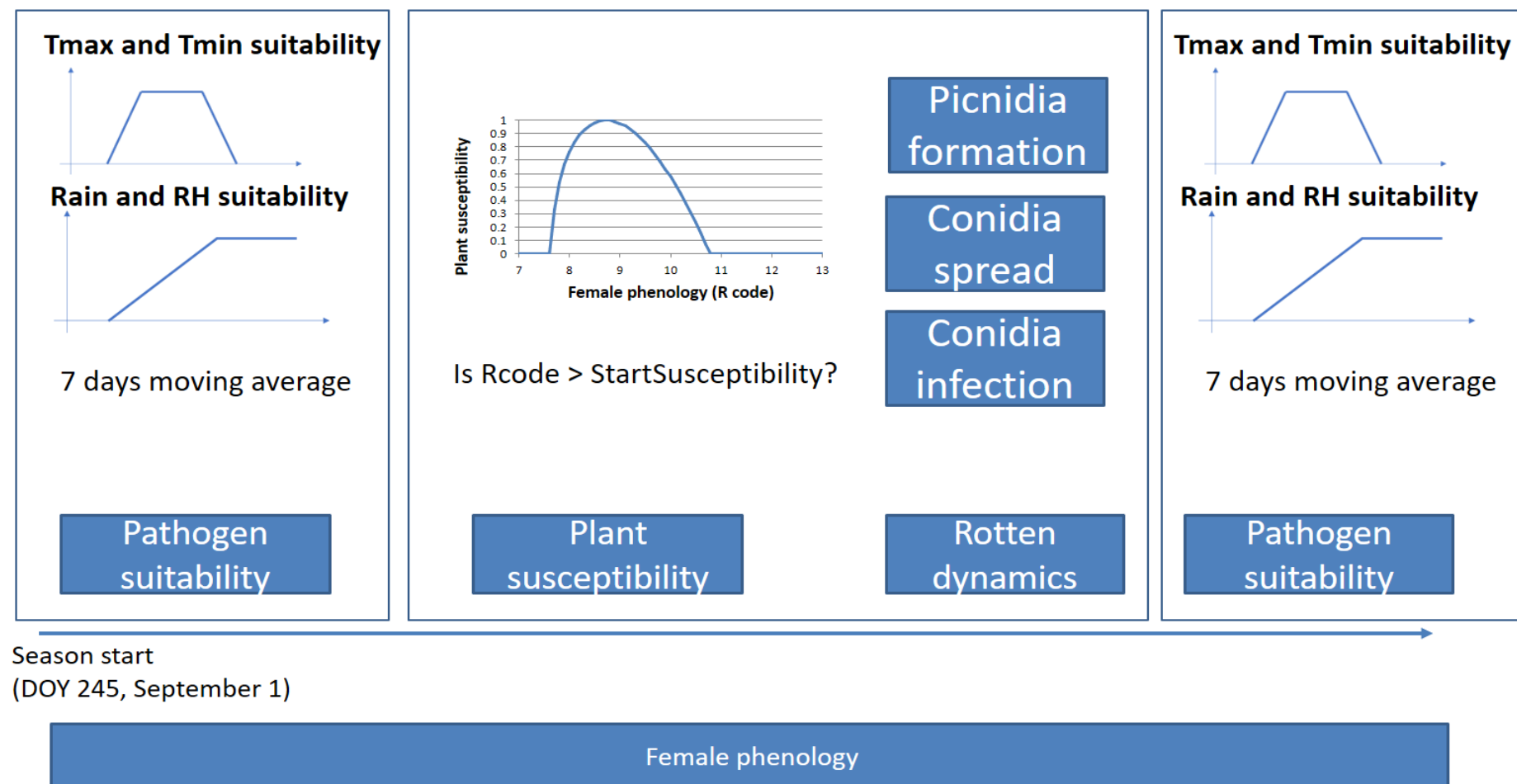
- Insect pests and diseases affect the quantitative and qualitative aspects of hazelnut yield, MatHiLDE project.
- Demand for decision support systems to optimize chemical control and forecasting activities to provide early yield predictions.

All kinds of defects, including brown spotted or moldy nut kernels.

Uncertain causal agent
Diaporthe spp (dominant)

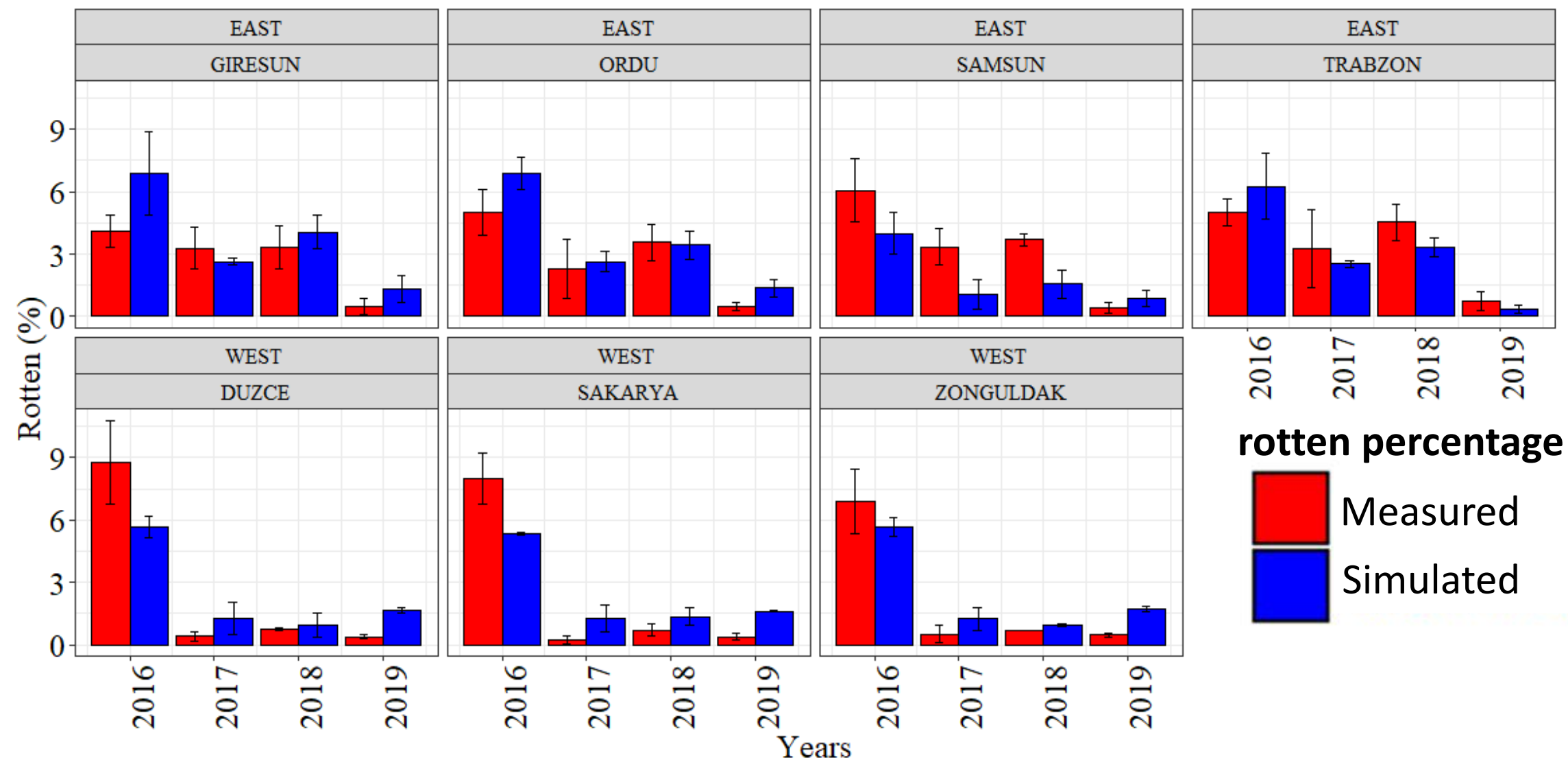


Model workflow



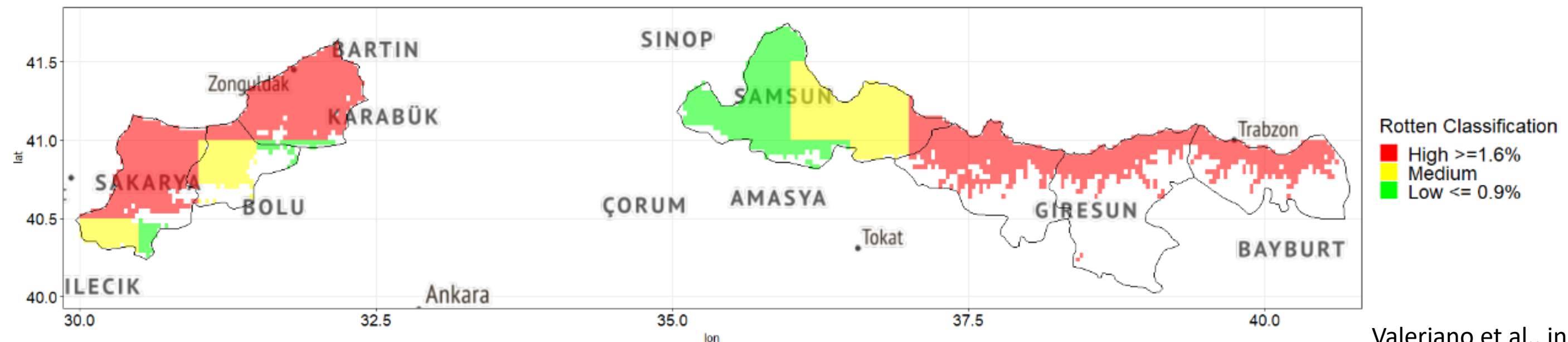
Use cases: rotten hazelnuts

- Four years of data (2016-2019) with around 100 sampling points per year on the main Turkish hazelnut producing areas (smartapp under development).
- Automatic calibration on 22 locations moving parameters within literature ranges.
- Model evaluation with independent subsets of observations.



Use cases: rotten hazelnuts

- Monthly bulletins since February 2020 using past weather data as “possible futures”.
- Model executed on NASA Power grids and on *in-situ* weather stations.
- Post-processing of model outputs with machine learning algorithms.
- Clear overestimation of field data in 2020 → problems to solve, back to model formalization.
- Sensitivity analysis under development to inspect model behaviour



Valeriano et al., in preparation

-

```

graph TD
    Start[Start Blast Simulation] --> Onset[Disease Onset]
    Onset --> Infection[Infection]
    Infection --> Latency[Latency]
    Latency --> Sporulation[Sporulation]
    Sporulation --> Dispersal[Dispersal]
    Dispersal --> Severity[Leaf Blast Severity/  
Panicle Blast Incidence]

    subgraph Inputs
        direction TB
        A1[Air Temperature]
        A2[Relative Humidity]
        A3[Air Temperature]
        A4[Min., Opt. and Max. Temperature for Infection/Latency Period]
        A5[Min., Opt. and Max. Temperature for Sporulation Period]
        A6[Min. Relative Humidity for Sporulation]
    end

    subgraph Functions
        direction TB
        F1([Hydro-thermal time])
        F2([Wetness Duration Requirement])
        F3([Air Temperature Response Function])
        F4([Sporulation Efficiency])
        F5([Latency Period])
        F6([Sporulation Dispersal Efficiency])
    end

    subgraph Parameters
        direction TB
        P1[Min. Temperature for Disease Onset]
        P2[Threshold for Leaf Wetness]
        P3[Min. and Max. Leaf Wetness Duration for Infection]
        P4[Critical Interruption of Dry Period D50]
        P5[Min. Latency Duration]
        P6[Wind Dispersal Parameters]
        P7[Rainfall causing 50% spore dispersal]
        P8[Max. Severity/Incidence Increase Hourly]
        P9[Susceptible Coefficient]
    end

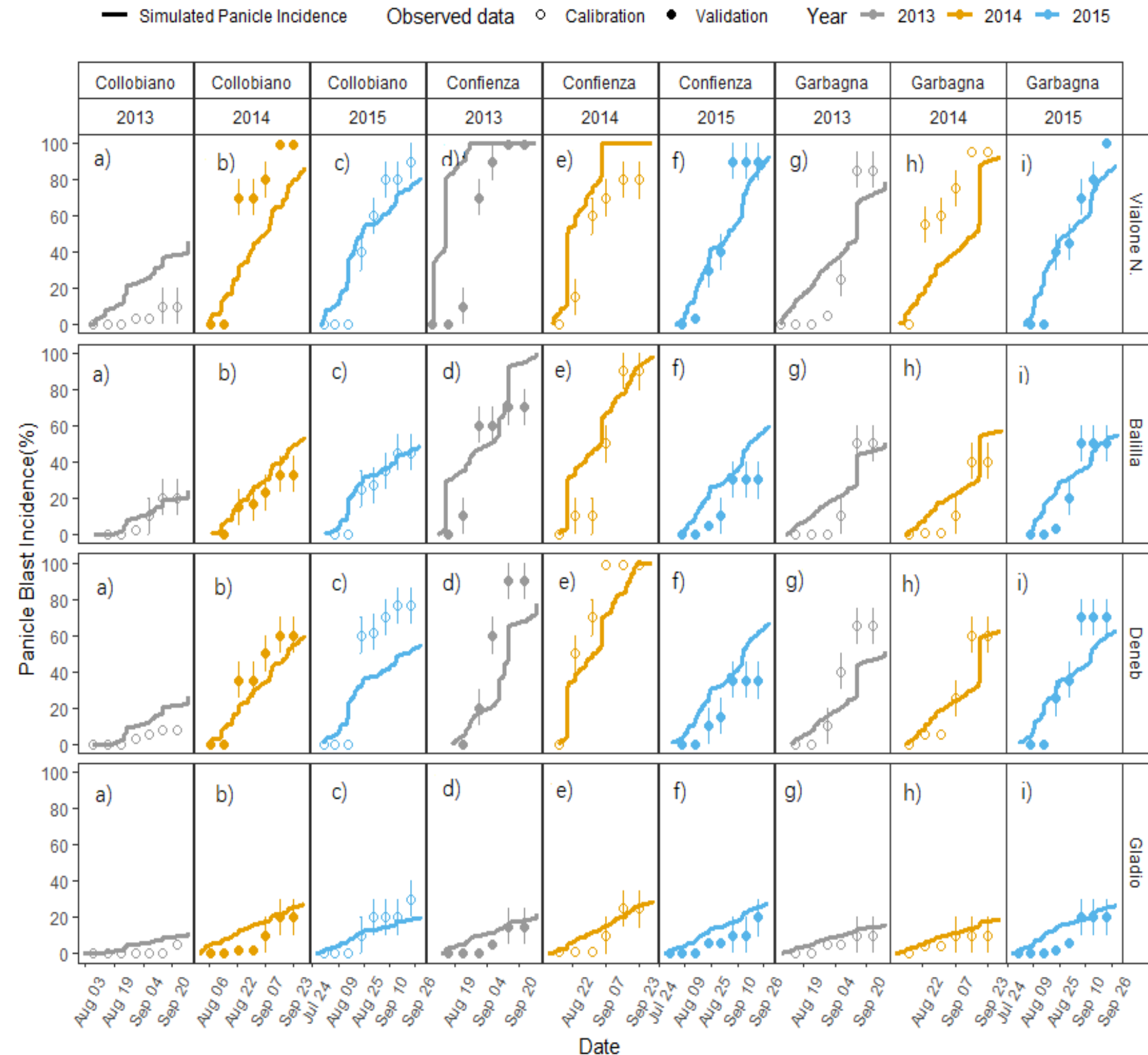
    A1 --> F1
    A2 --> F1
    A3 --> F3
    A4 --> F3
    A5 --> F4
    A6 --> F4
    F1 --> Onset
    F2 --> Infection
    F3 --> Infection
    F3 --> Latency
    F3 --> Sporulation
    F3 --> Severity
    F4 --> Sporulation
    F4 --> Severity
    F5 --> Latency
    F6 --> Dispersal
    P1 --> F1
    P2 --> F1
    P3 --> F2
    P4 --> F2
    P5 --> F5
    P6 --> F6
    P7 --> F6
    P8 --> P9
    P9 --> Severity
  
```

The flowchart illustrates the progression of leaf blast simulation, starting from 'Start Blast Simulation' and moving through 'Disease Onset', 'Infection', 'Latency', 'Sporulation', and 'Dispersal' to 'Leaf Blast Severity/Panicle Blast Incidence'. The process is influenced by various environmental and physiological parameters, which are categorized into three main groups: Inputs, Functions, and Parameters. The 'Inputs' group includes 'Air Temperature' and 'Relative Humidity' for 'Hydro-thermal time', 'Air Temperature' for 'Air Temperature Response Function', and 'Min., Opt. and Max. Temperature for Infection/Latency Period' and 'Min., Opt. and Max. Temperature for Sporulation Period' for 'Sporulation Efficiency'. The 'Functions' group includes 'Hydro-thermal time', 'Wetness Duration Requirement', 'Air Temperature Response Function', 'Sporulation Efficiency', 'Latency Period', and 'Sporulation Dispersal Efficiency'. The 'Parameters' group includes 'Min. Temperature for Disease Onset', 'Threshold for Leaf Wetness', 'Min. and Max. Leaf Wetness Duration for Infection', 'Critical Interruption of Dry Period (D50)', 'Min. Latency Duration', 'Wind Dispersal Parameters', 'Rainfall causing 50% spore dispersal', 'Max. Severity/Incidence Increase Hourly', and 'Susceptible Coefficient'. The 'Parameters' group is further divided into 'Wind Dispersal Parameters' and 'Rainfall causing 50% spore dispersal', which are influenced by 'Wind Speed' and 'Rainfall'.

Use cases: rice blast in Italy

■ Field experimental data on four Italian varieties at variable resistance

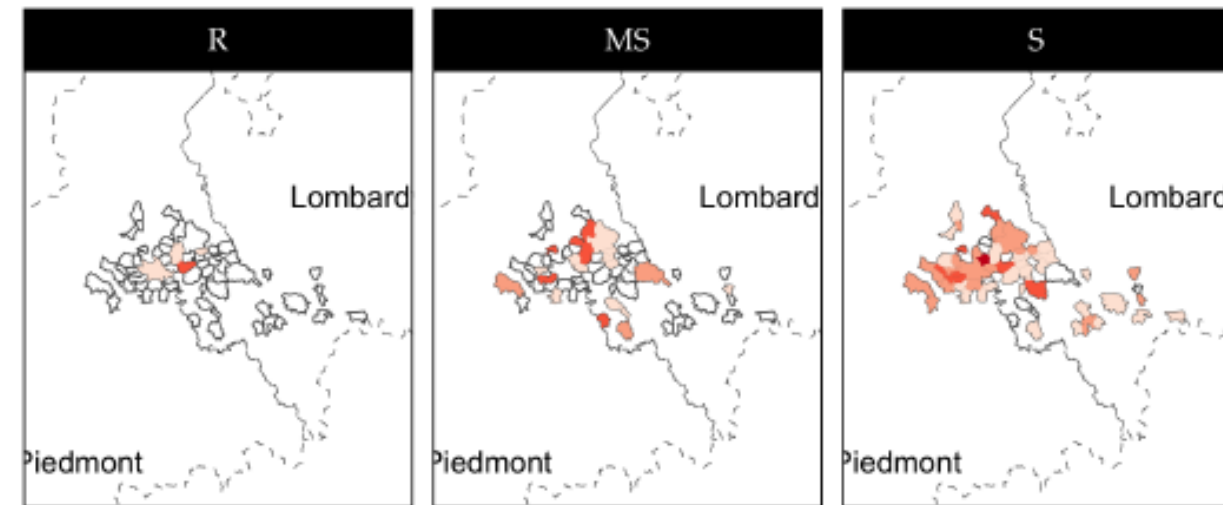
■ Same model parameterized for leaf and panicle blast



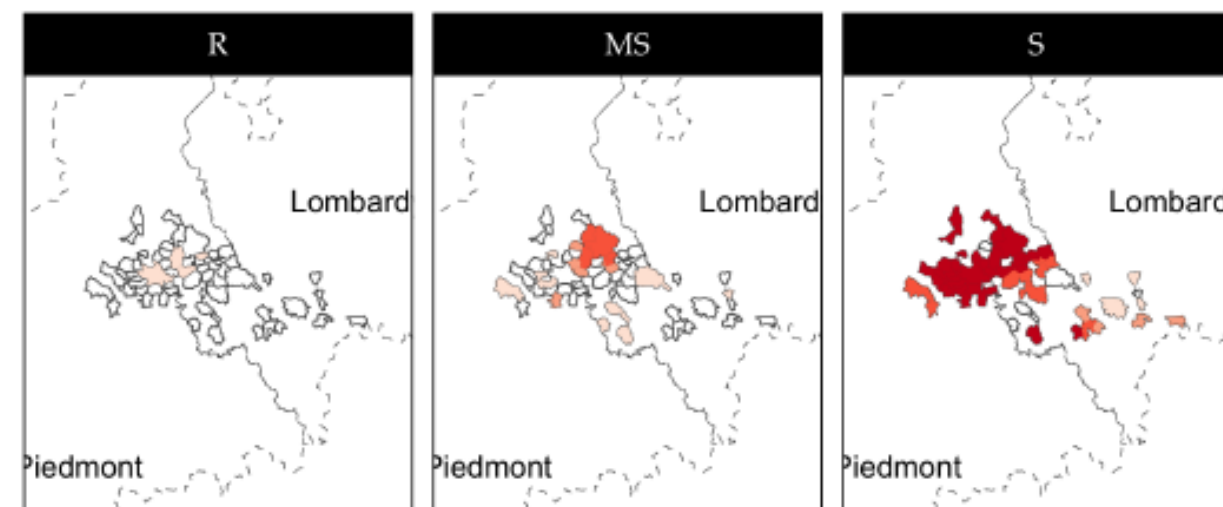
Use cases: rice blast in Italy

- Field experimental data on four Italian varieties at variable resistance
- Same model parameterized for leaf and panicle blast
- Further model evaluation with data at municipality level

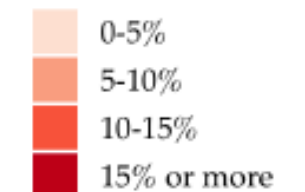
A
Observed



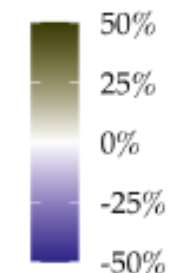
B
Predicted



Panicle Blast Incidence

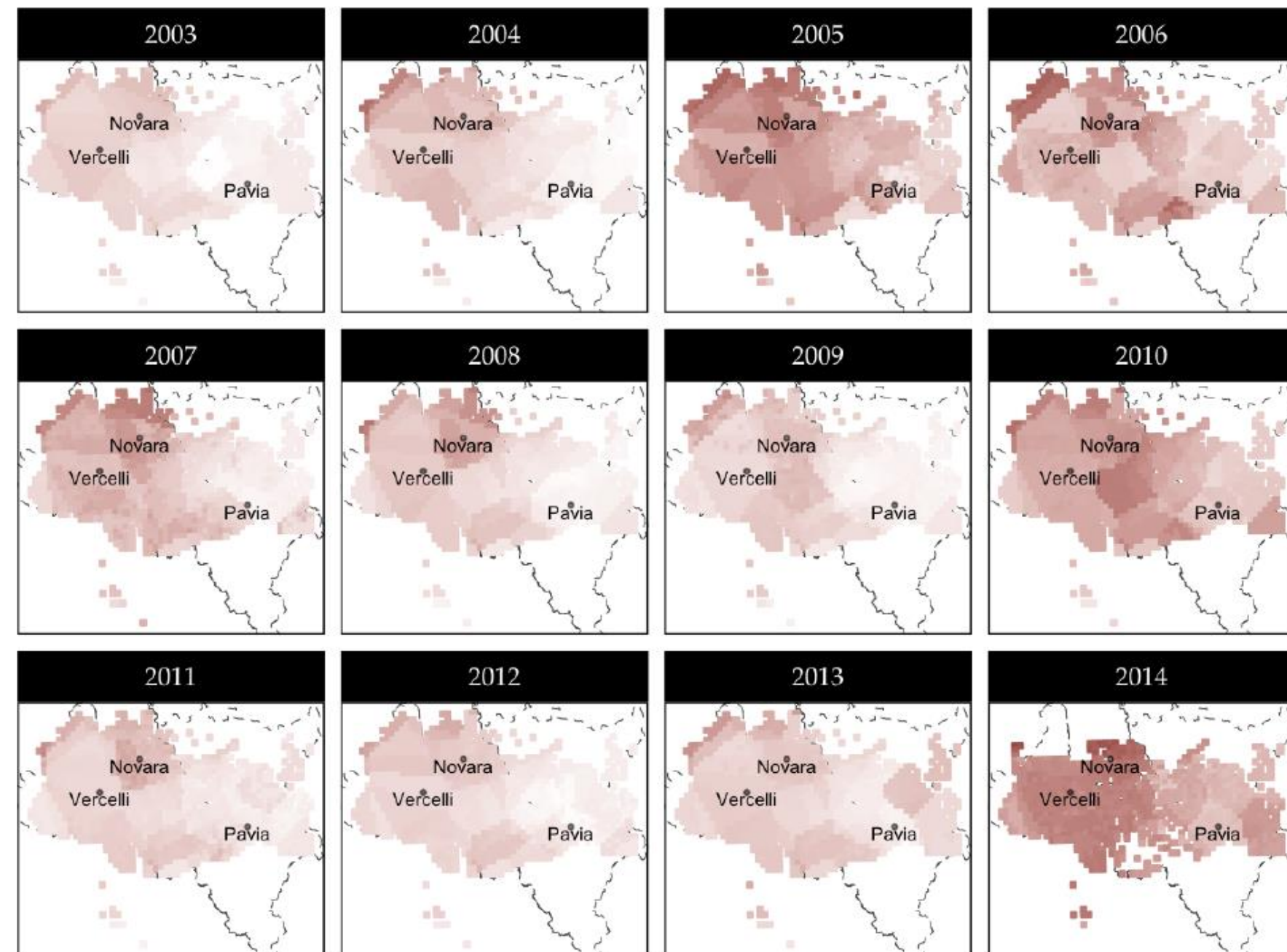


Difference



Use cases: rice blast in Italy

- Field experimental data on four Italian varieties at variable resistance
- Same model parameterized for leaf and panicle blast
- Further model evaluation with data at municipality level
- Model application in the main Italian rice district



Panicle Blast Incidence

0%

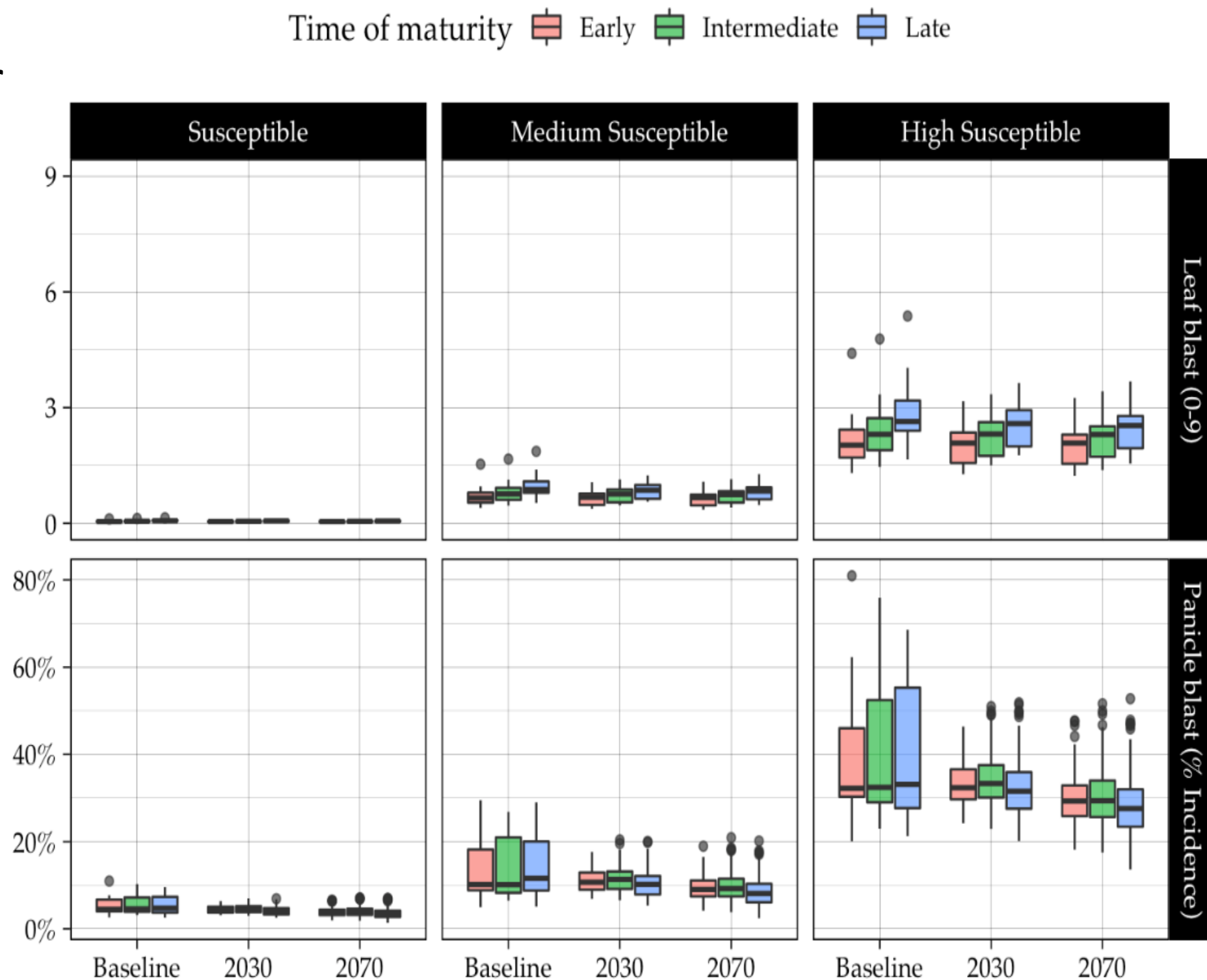
25%

50%

75%

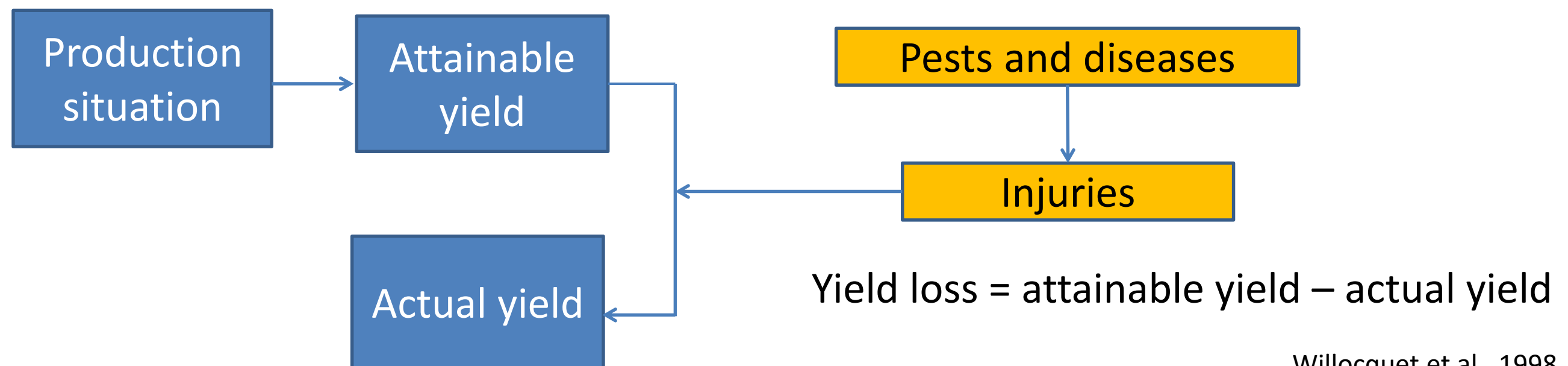
Use cases: rice blast in Italy

- Field experimental data on four Italian varieties at variable resistance.
- Same model parameterized for leaf and panicle blast.
- Further model evaluation with data at municipality level.
- Model application in the main Italian rice district.
- Simulations under climate change scenarios.



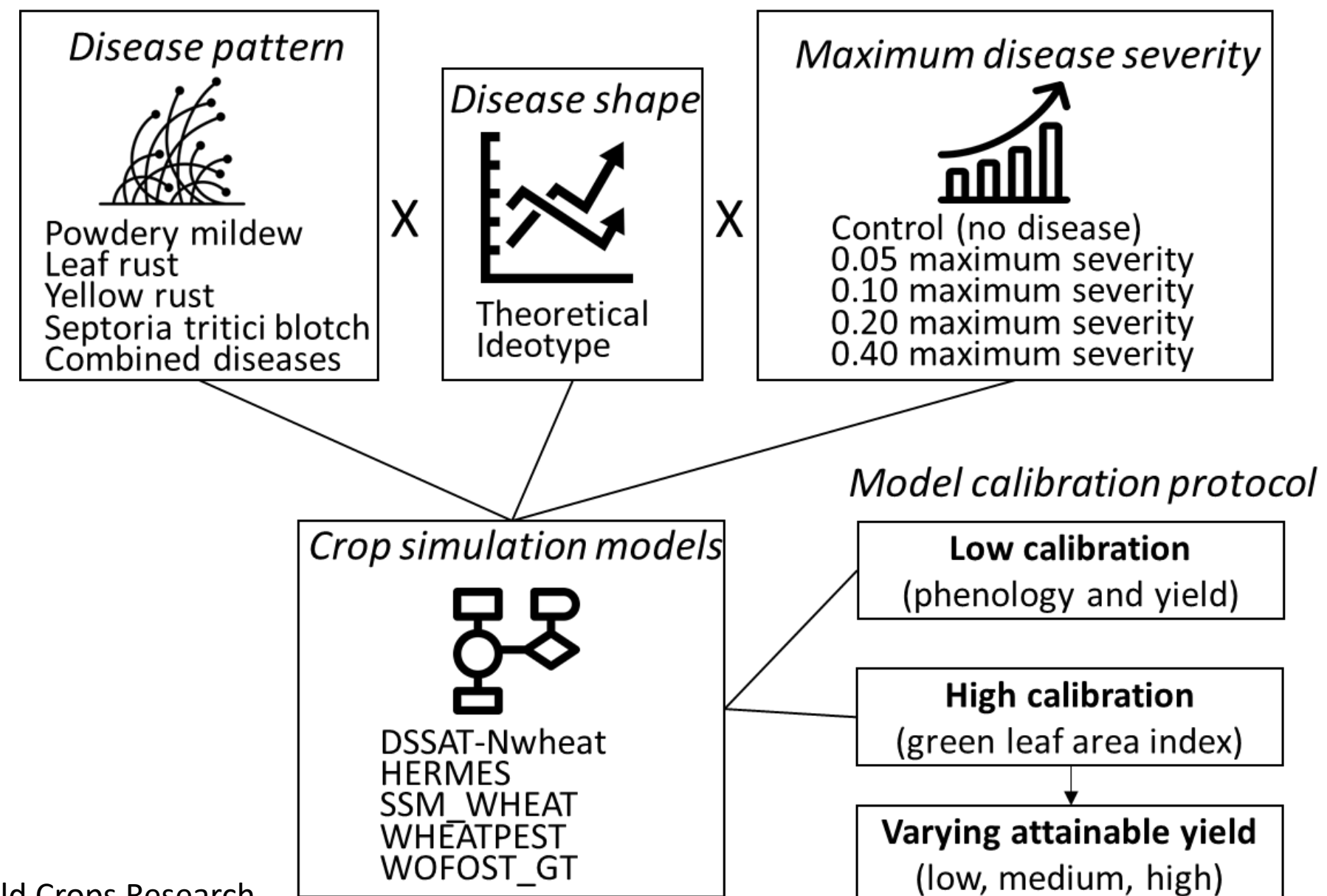
Incorporating damage mechanisms into crop models

- The **dynamic linkage** between **disease and pest injuries** and the **host crop** is realized through **coupling points** between **pest/disease and crop models**.
- **Seven pest and disease damage mechanisms** were identified (Boote et al., 1983, i.e., light stealer, leaf senescence accelerator, tissue consumer, stand reducer, photosynthetic rate reducer, turgor reducer and assimilate sappers).



Incorporating damage mechanisms into crop models

- The simulation experiment design: WHEATPEST (Willoquet et al., 2008) reference model to incorporate damage mechanisms into several wheat models.
- Simulation protocol to identify the main sources of uncertainty in yield loss modelling.
- Simulations performed at two calibration levels (i.e., with and without leaf area index data).



major revision on Field Crops Research

Main outcomes

- 3-way ANOVAs were performed to investigate the effect of the models, diseases, and disease severity levels, separately for each combination of calibration level and injury driver.
- F values are highest for "disease level" factor, which has the highest effect on yield loss, intermediate for "model" and smallest for "disease". F values are reduced when model calibration is higher.

| | High calibration | | Low calibration | |
|---------------|--------------------|--------------------|--------------------|--------------------|
| | Ideotype injury | Theoretical injury | Ideotype injury | theoretical injury |
| Model | 30 ^{***} | 27 ^{***} | 41 ^{***} | 45 ^{***} |
| Disease | 4 ^{***} | 2 ^{ns} | 4 ^{***} | 2 ^{ns} |
| Disease level | 154 ^{***} | 296 ^{***} | 120 ^{***} | 228 ^{***} |

Perspectives

Weather databases

Historical series (1961-2017) for models calibration/evaluation



Agrometeorological stations

Real time data for disease forecasting



Remote sensing

Correlation with disease data, large area monitoring



Weather forecasts

Short-term data (3-7 days)



Coupled crop-diseases models



We are focusing on diseases of grape, rice and olive

Models calibration
evaluation



Past observations
Regional services



Smart data field survey
Real time data (MIRA)



Custom dashboard
Field samplings and models
projections



**Model outputs to regional
services**

A common workflow for the three case studies



| Need | Action | Data and skills |
|--|--|---|
| Identify the priorities in pests/disease modelling | Stakeholders involvement to characterize urgent needs | Surveys, meetings, literature search, working group, research projects. |
| Acquire knowledge on the target pathosystem | Literature search on the pathogen biological cycle and on the factors influencing epidemics | Laboratory studies where the components of the pathogens cycle are studied as a function of climate |
| Formalize the models to simulate the key epidemiological processes | Parameterization (existing models) or development of models for specific processes, and their coupling | Programming skills to implement models, adequate software frameworks, sensitivity analyses to test models behaviour |
| Calibration and evaluation of models with ground truth | Test of the models accuracy in reproducing reference data for the target pathosystems | Agro-meteorological data, reference disease data, automatic calibration tools to optimize models performances |
| Large area model application for monitoring or forecasting | Link with georeferenced databases, execute simulations and extract information | Public/private databases, GIS, IT skills to automatize models execution, data analysis to elaborate model outputs. |

THANK YOU FOR YOUR ATTENTION

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