

Simulating the spread of flying insects using HTC

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Yan Boulanger, Canadian Forest Service

Philip A. Townsend, UW–Madison F&WE

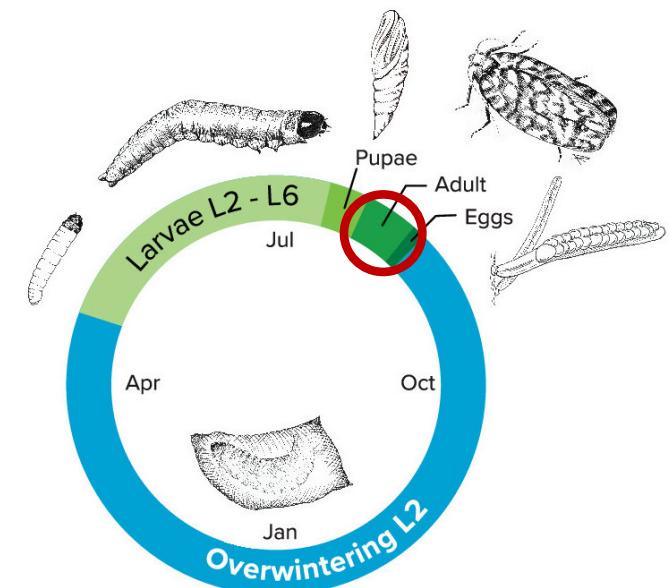
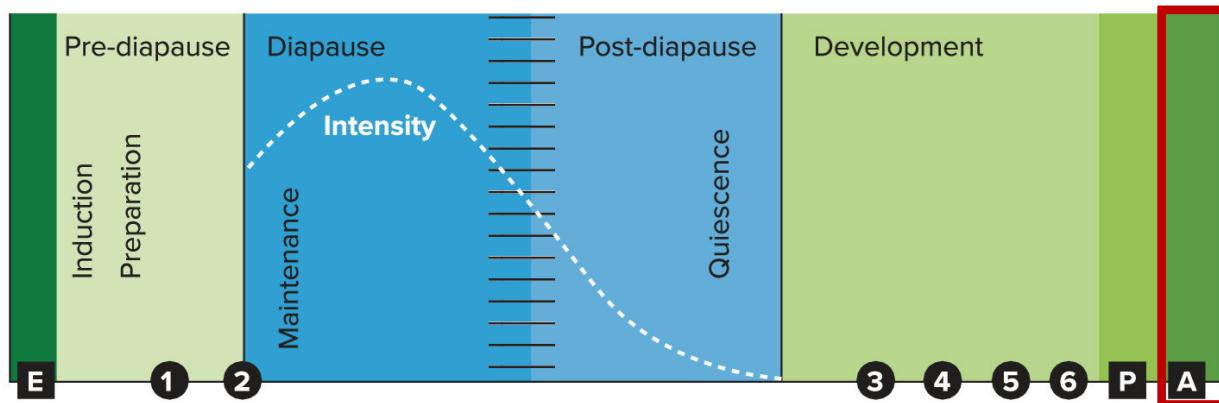


US-Canadian Forest Health Initiative
"One Continent, One Forest, One Threat"



Multidisciplinary project: modeling dispersal of SBW moths

Eastern spruce budworm (SBW) in boreal spruce–fir forests
life cycle, defoliation, aerial dispersal, oviposition

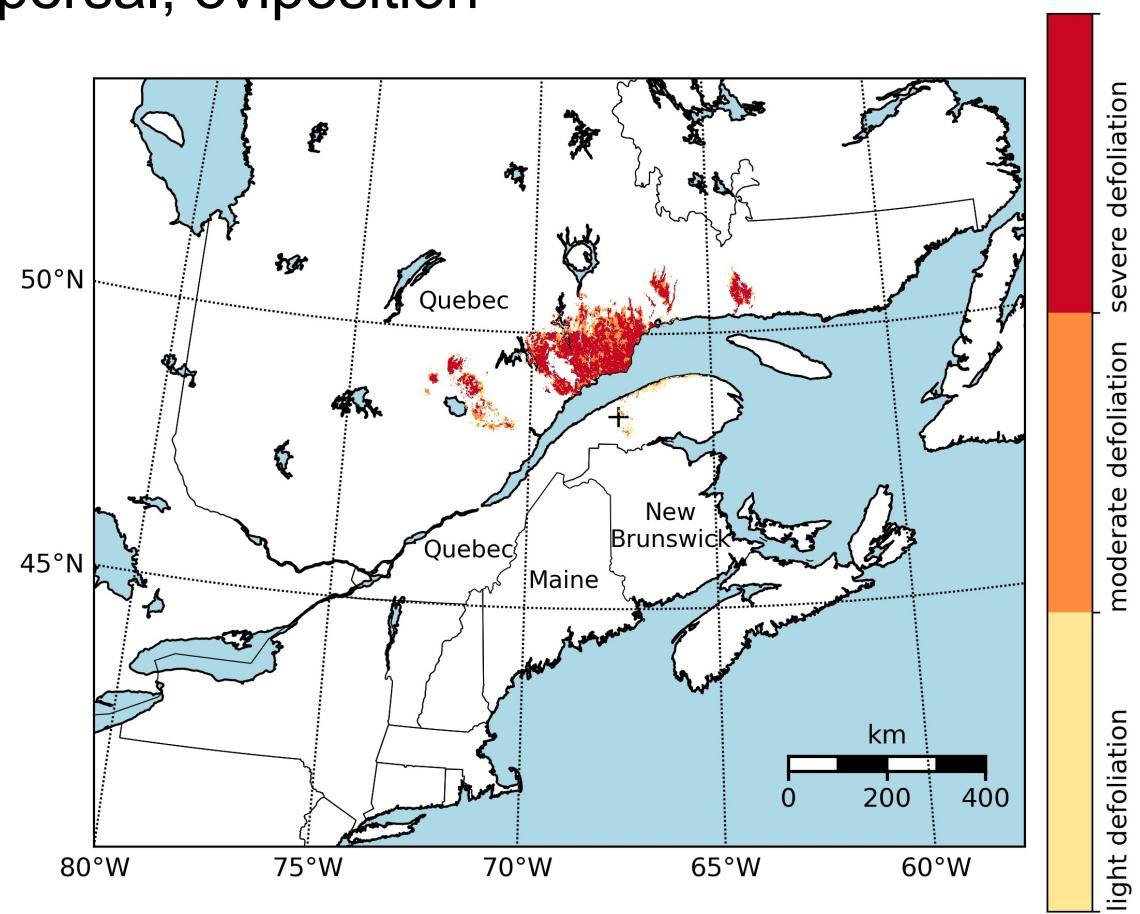


Development cycle graphics
adapted from
Marshall and Roe (2021, *Physiology*)

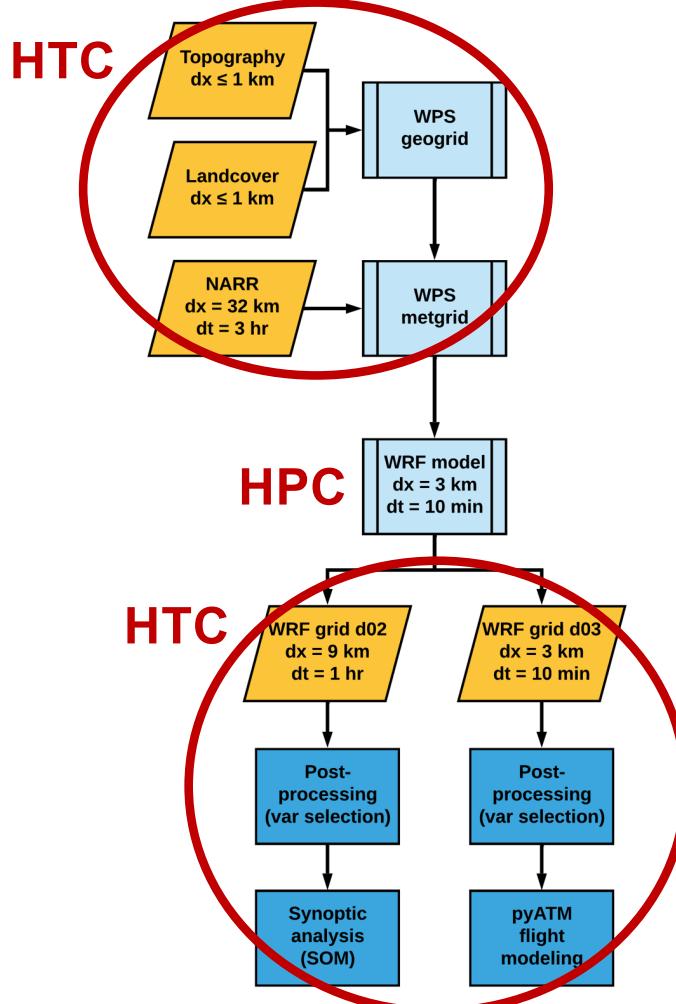
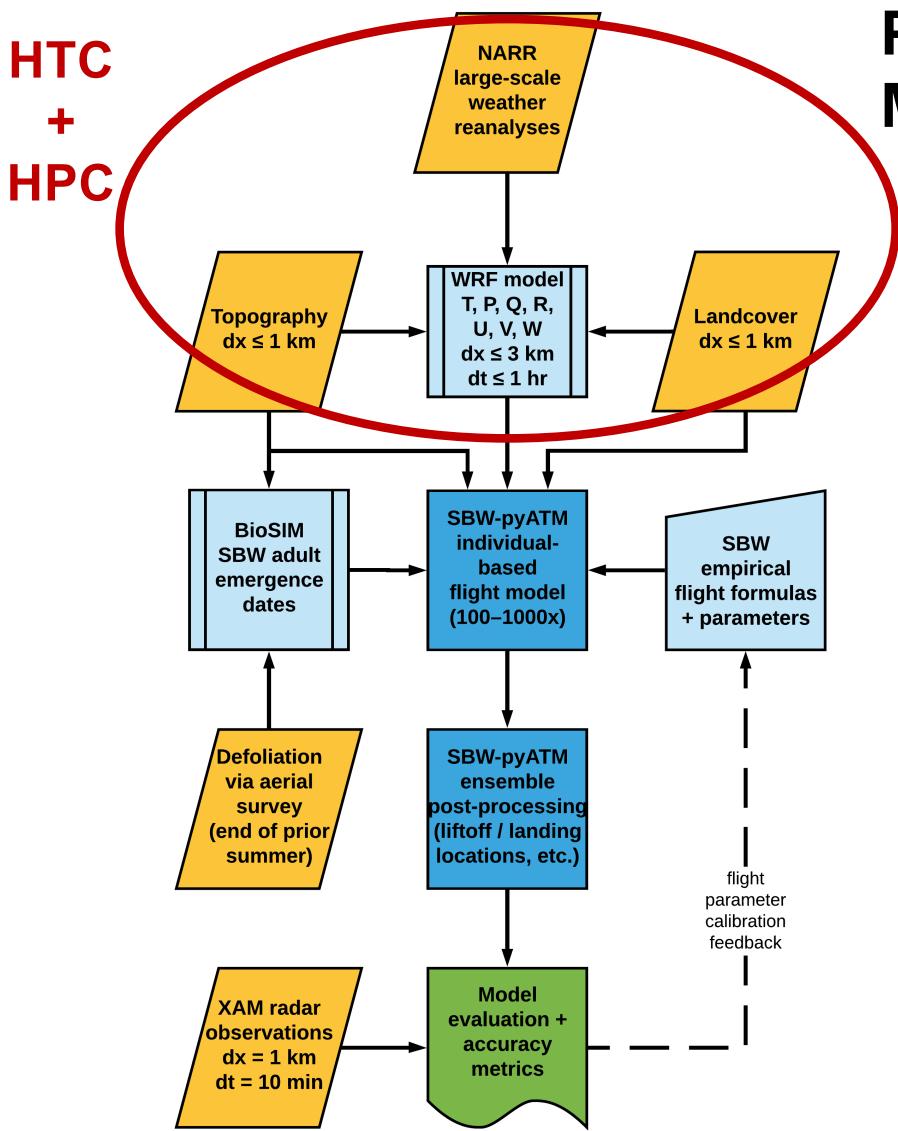
Multidisciplinary project: modeling dispersal of SBW moths

Eastern spruce budworm (SBW) in boreal spruce–fir forests
life cycle, defoliation, aerial dispersal, oviposition

Focus on Quebec +
New Brunswick +
Maine



Python-based Atmospheric Transport Model (pyATM) for SBW moths



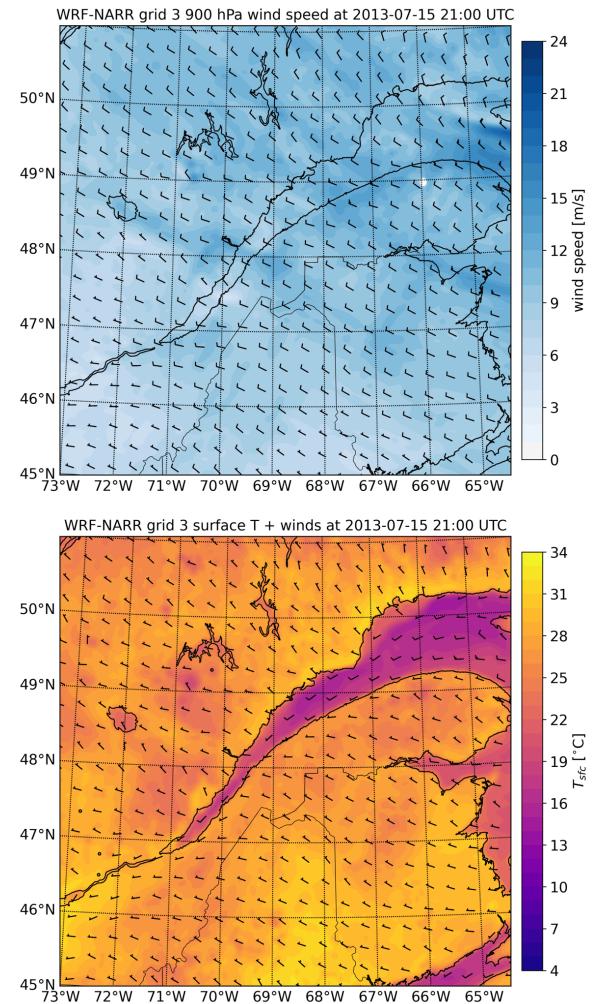
Weather Research & Forecasting (WRF) Model v4.1

Pre-processing: 1 month at a time
single-processor, high-memory

Main WRF model: 1 day (30h) at a time
60+ processors, ~2GB memory each
separate days can run concurrently

Post-processing: 1 output file at a time
single-processor, distributed

HTCondor DAGMan script(s)
with mixed submission protocols
HTC → HPC → HTC

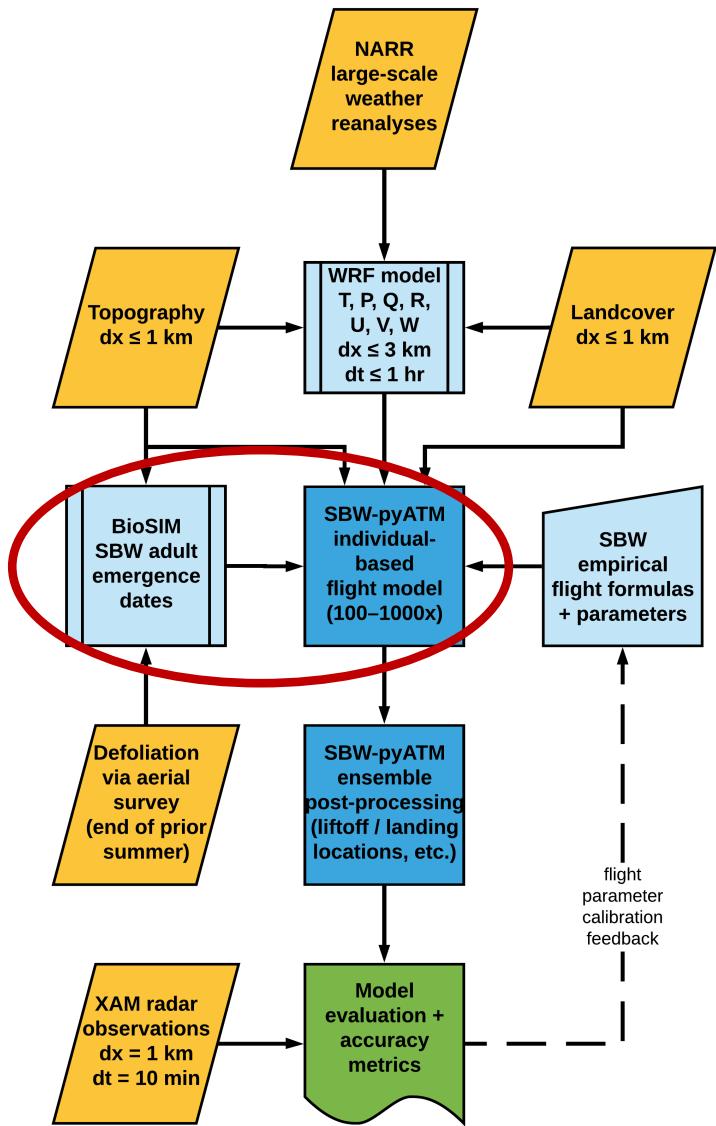


Python-based Atmospheric Transport Model (pyATM) for SBW moths

There is no way (yet) to count the SBW moth population, but say $\sim 10^{11}$

Simulate 10^3 random moths

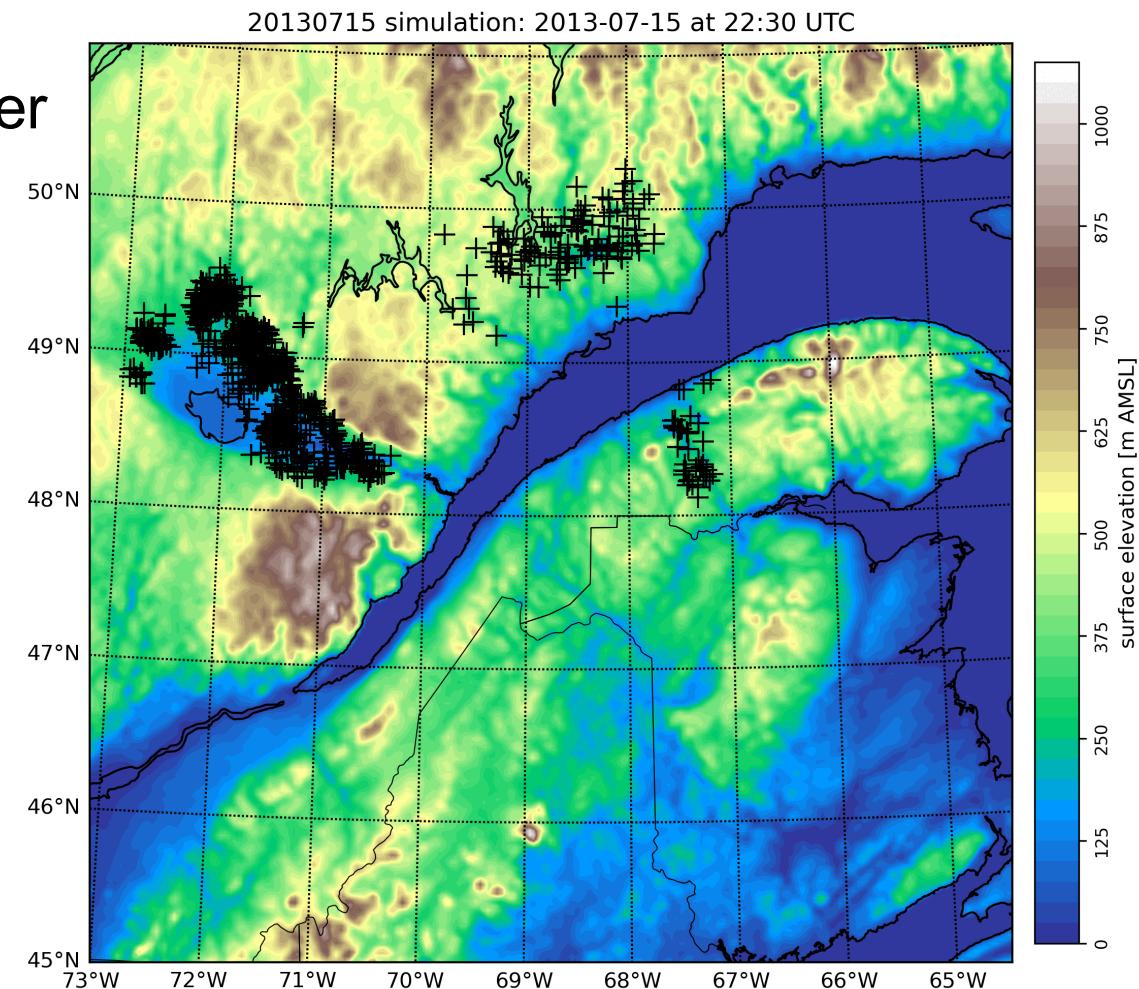
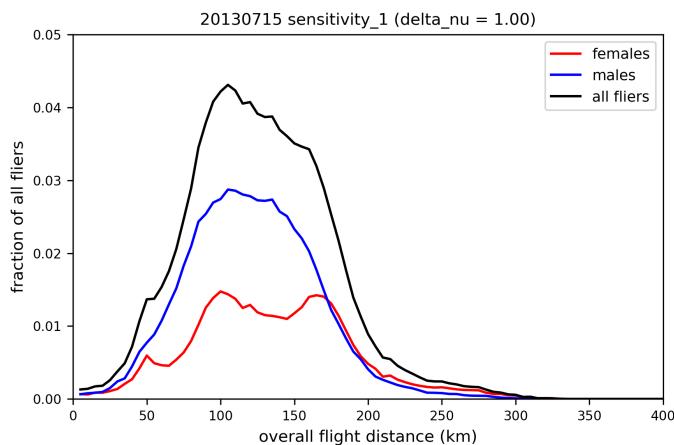
HTC

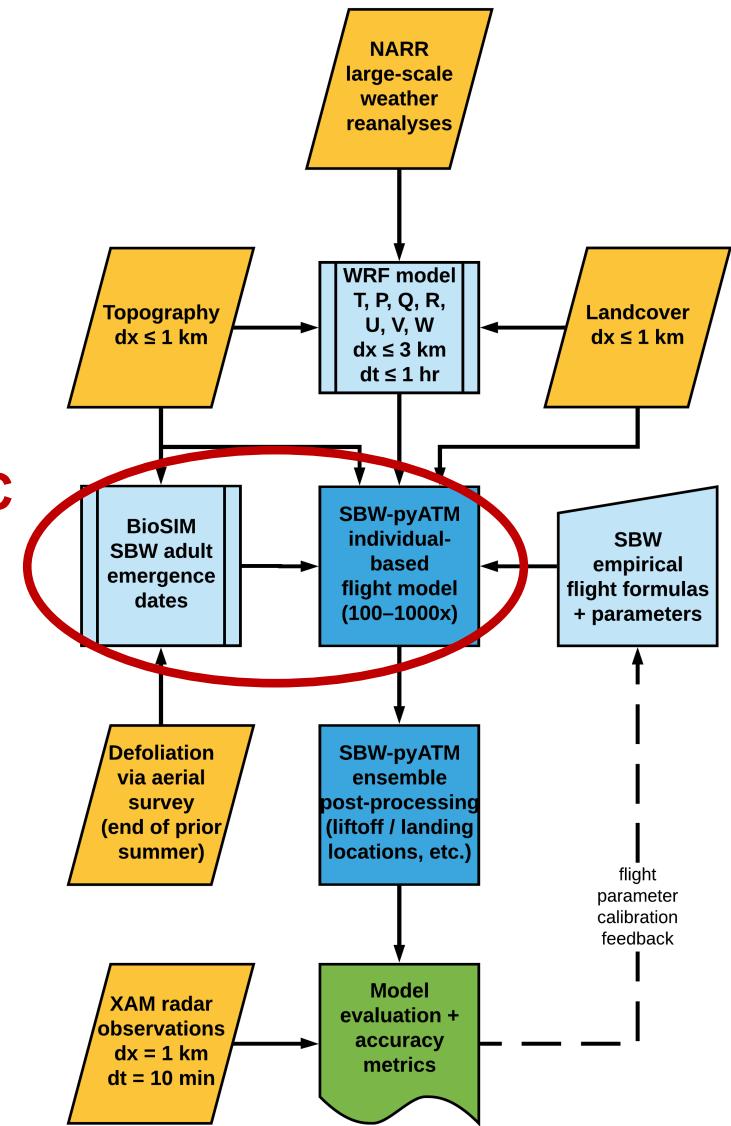


Python-based Atmospheric Transport Model (pyATM) for SBW moths

Adult dispersal is weather-driven,
occurs almost nightly in summer

M + F have diff. flight capabilities
→ flight altitudes + distances





Python-based Atmospheric Transport Model (pyATM) for SBW moths

There is no way (yet) to count the SBW moth population, but say $\sim 10^{11}$

Simulate 10^3 random moths
 \rightarrow 2–12 GB mem, 10–18 GB disk, 2–8 h

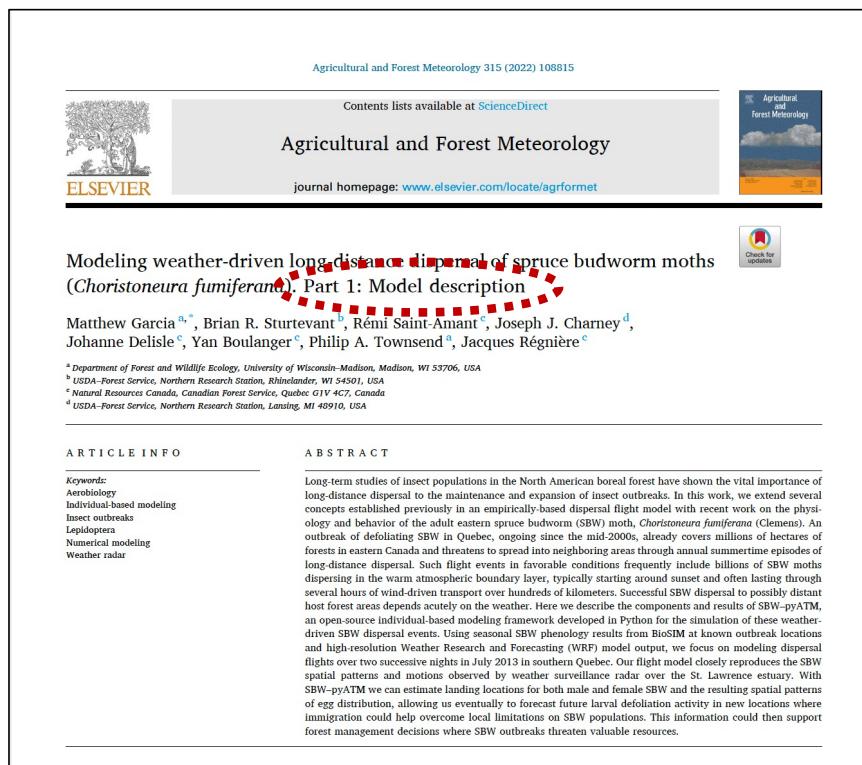
Consider a representative (?) sample
 $\rightarrow 10^6$ moths ($\sim 1/10^5 \times$ the pop. guess)

So, simulate 10^3 random moths 10^3 times
 \rightarrow Monte Carlo ensemble approach
 \rightarrow HTC with UW + OSG resources

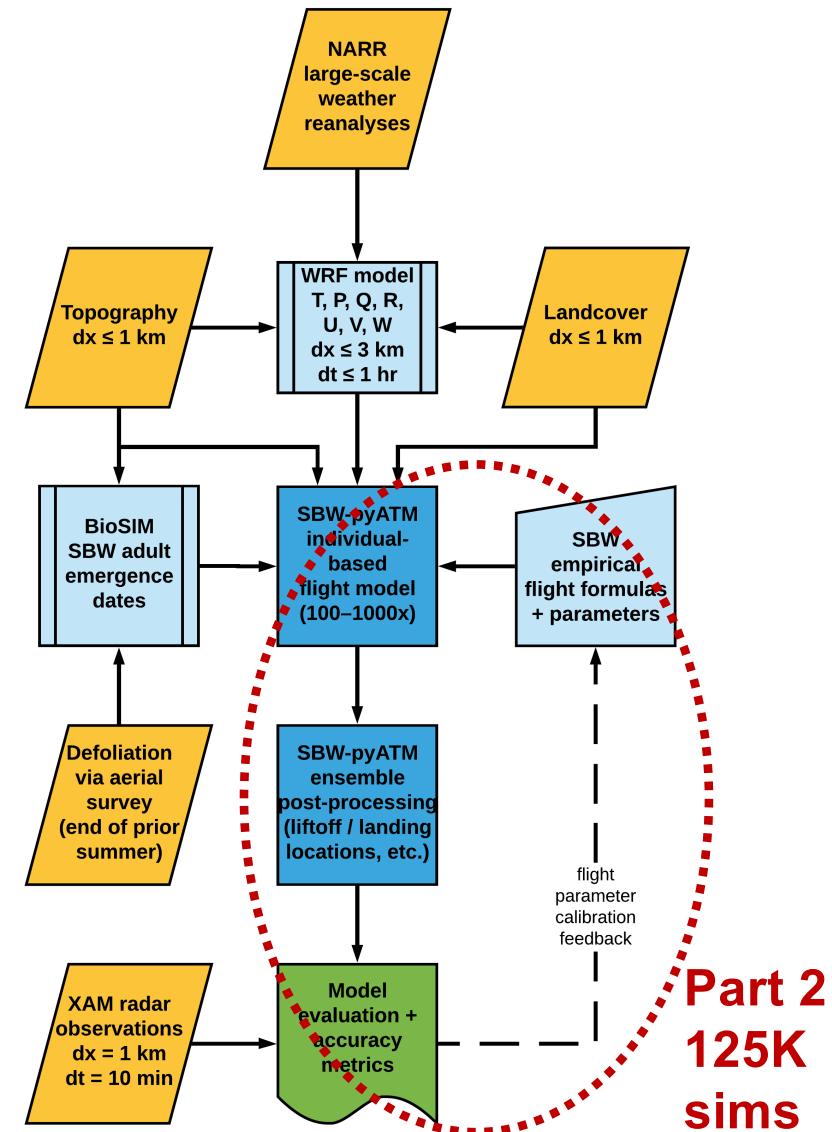
Also: calibration w/ radar, uncertain parameters, sequential nights, etc.

Python-based Atmospheric Transport Model (pyATM) for SBW moths

Part 1: Model description (pub'd Jan 2022)



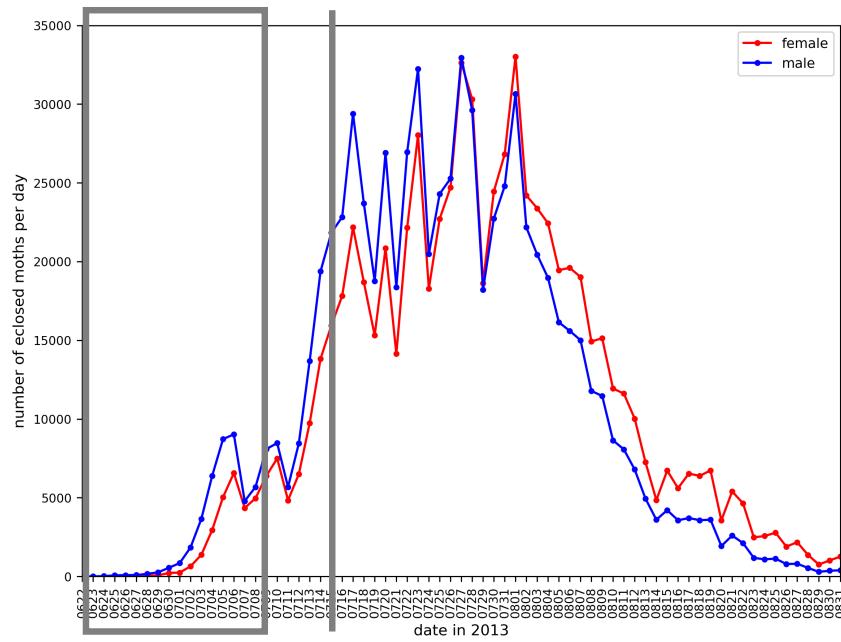
doi: 10.1016/j.agrformet.2022.108815



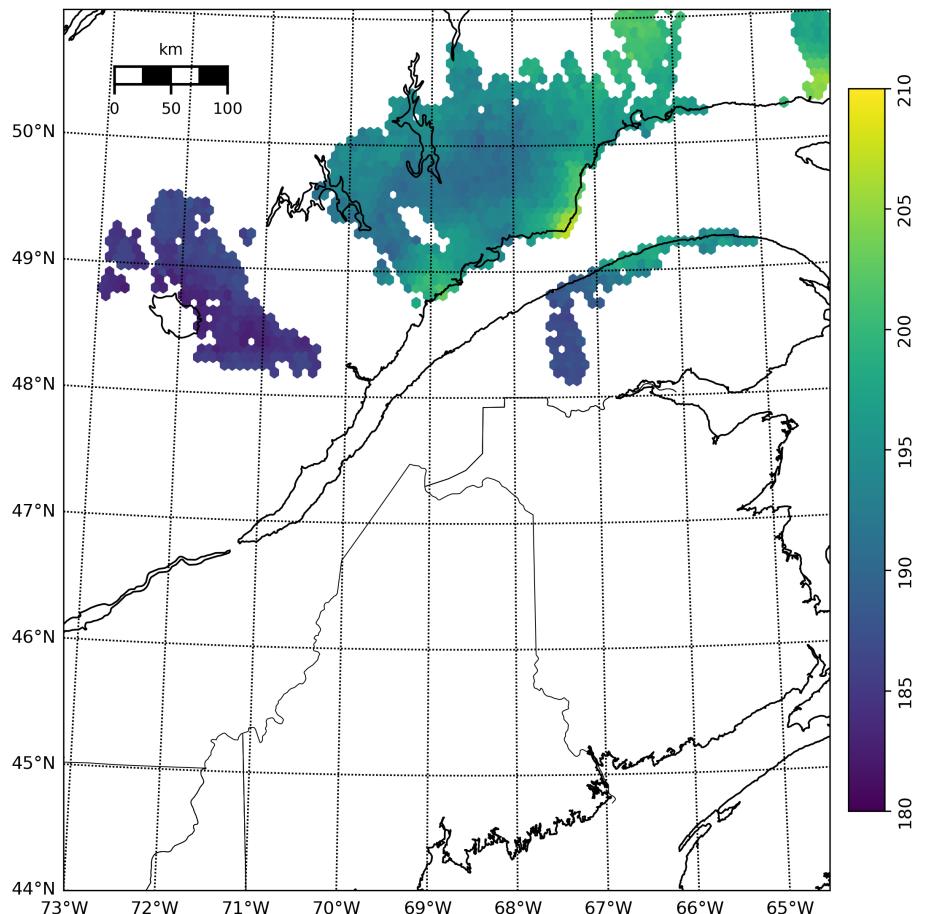
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Each night's cohort includes “new”
moths *and* those that survived
the previous night(s)



2013 BioSIM median DOY for adult SBW eclosion

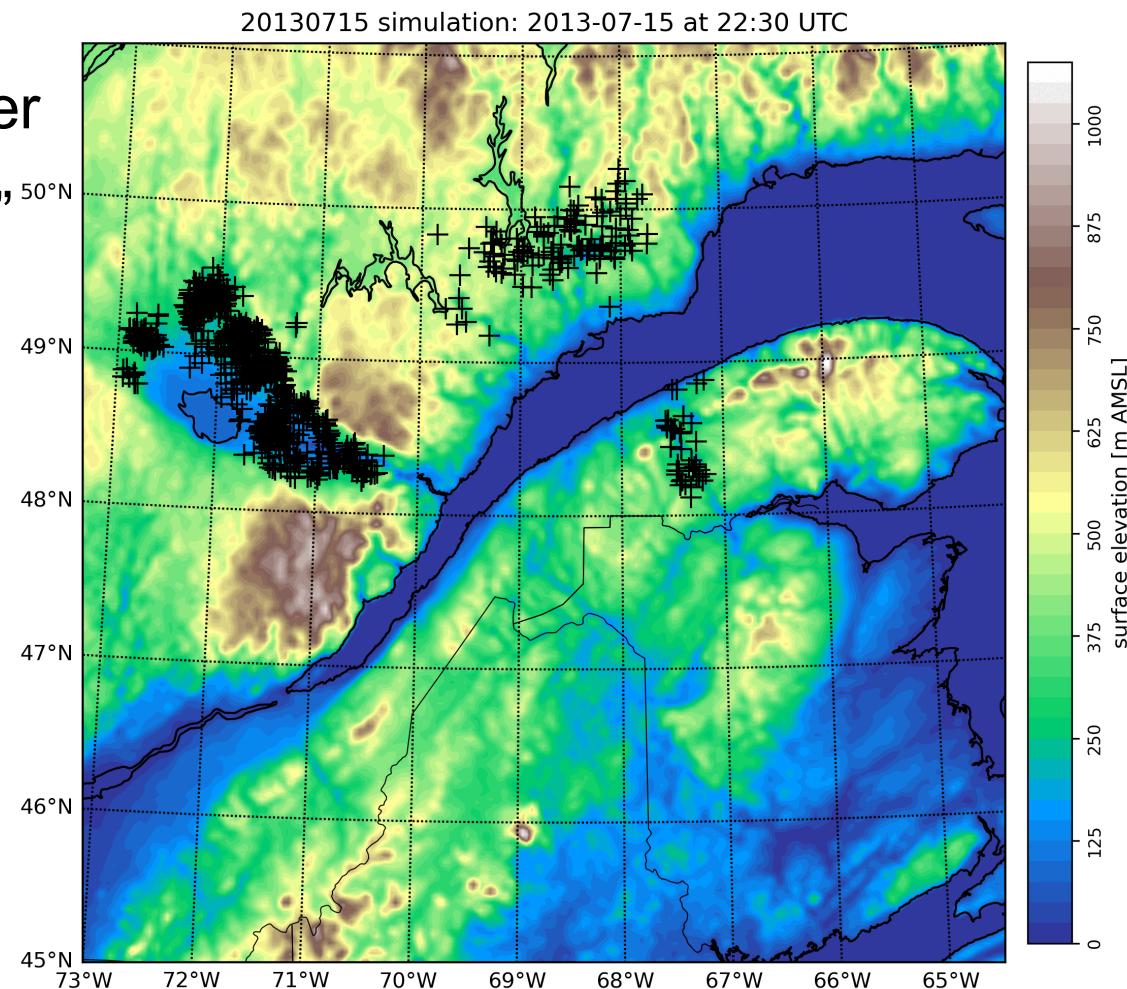


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F lay eggs on host trees, poss.
at each new landing location
→ daily changes in weight
and flight capability



Python-based Atmospheric Transport Model (pyATM) for SBW moths

Stochastic simulations

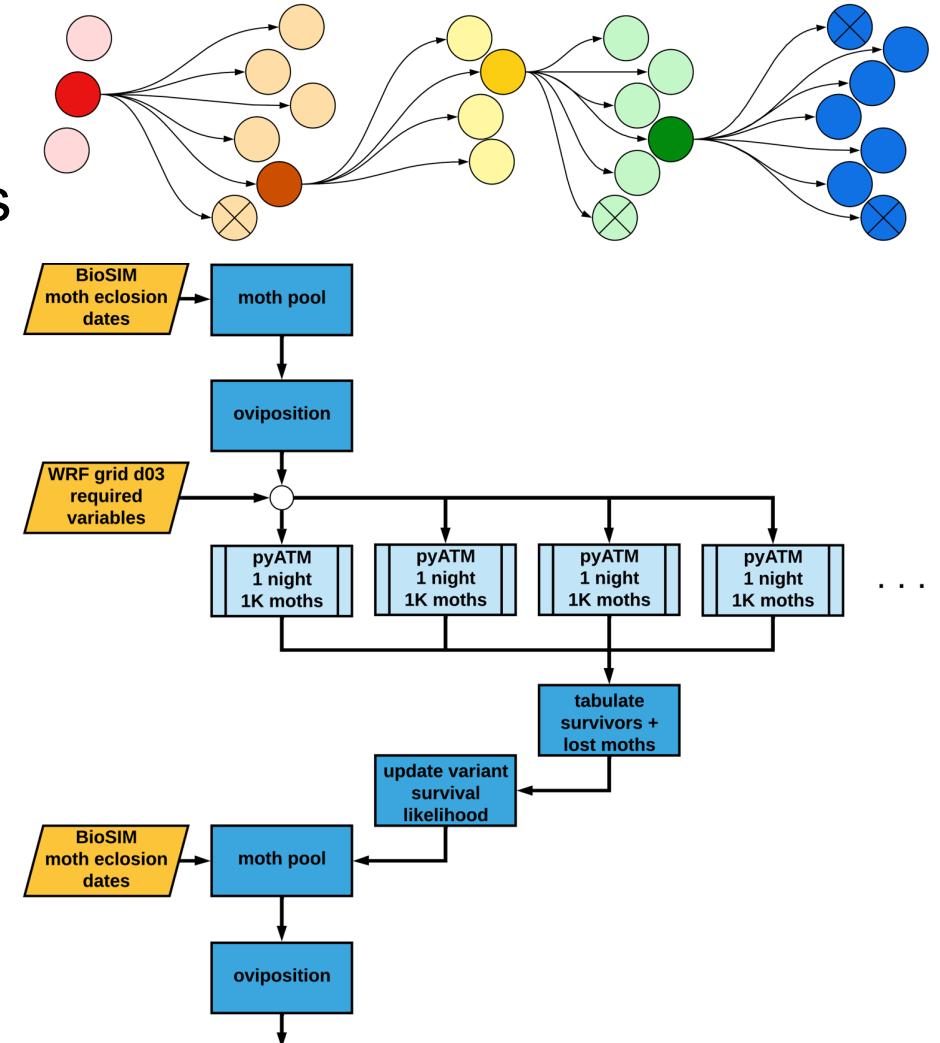
- random subsets of available moths
- liftoff times vary with local conditions
- weather, esp. near the surface (BL)

→ Monte Carlo ensemble of simulations

Sequential nightly simulations

- accounting variety of prior results
- compounded timelines + variants

→ Markov chain Monte Carlo process
with Bayesian likelihood updating



Python-based Atmospheric Transport Model (pyATM) for SBW moths

```
# 2013_sequential_dag.sub with daily accounting of moth population and dynamic generation of flight modeling ensembles

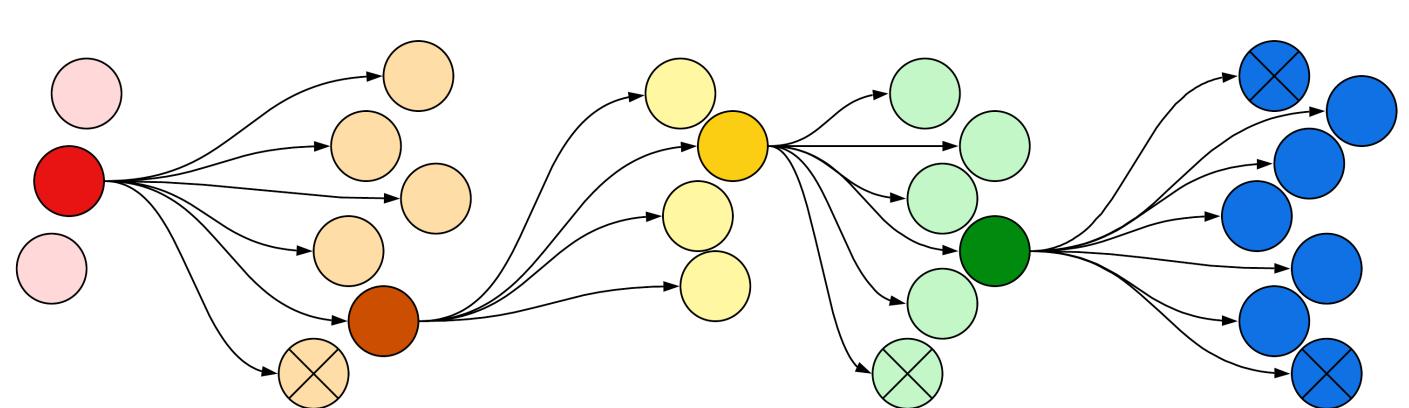
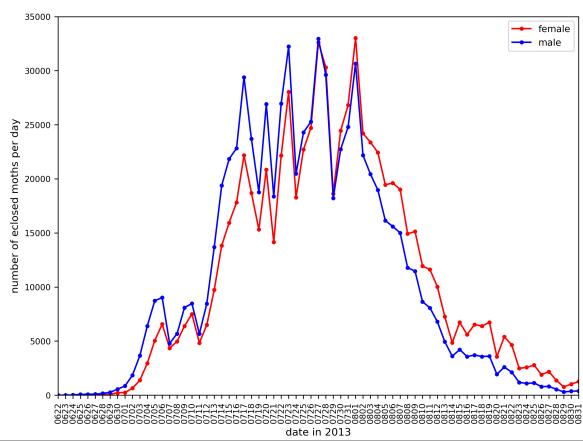
JOB SBM_20130624 ATM_WRF-NARR_preprocessing_first.sub
VARS SBM_20130624 grid="d03" begin_date="20130624" begin_year="2013"
SCRIPT POST SBM_20130624 ATM_WRF-NARR_preprocessing_post.sh d03 20130624 20130625

SUBDAG EXTERNAL ATM_20130624 ATM_WRF-NARR_d03_20130624_subdag.sub
PARENT SBM_20130624 CHILD ATM_20130624

JOB SBM_20130625 ATM_WRF-NARR_preprocessing.sub
VARS SBM_20130625 grid="d03" prev_date="20130624" begin_date="20130625" begin_year="2013"
SCRIPT PRE SBM_20130625 ATM_WRF-NARR_preprocessing_pre.sh d03 20130624
SCRIPT POST SBM_20130625 ATM_WRF-NARR_preprocessing_post.sh d03 20130625 20130626
PARENT ATM_20130624 CHILD SBM_20130625

SUBDAG EXTERNAL ATM_20130625 ATM_WRF-NARR_d03_20130625_subdag.sub
PARENT SBM_20130625 CHILD ATM_20130625

...
```



Python-based Atmospheric Transport Model (pyATM) for SBW moths

```
# ATM_WRF-NARR_d03_20130624_subdag.sub using dynamic generation of flight modeling ensemble

JOB ATM_20130624 ATM_WRF-NARR.sub
VARS ATM_20130624 grid="d03" begin_date="20130624" end_date="20130625" nreps="10"
SCRIPT PRE ATM_20130624 ATM_WRF-NARR_pre.sh d03 20130624 20130625
SCRIPT POST ATM_20130624 ATM_WRF-NARR_post.sh d03 20130624 20130625
```

```
# ATM_WRF-NARR.sub template submit file using Flocking + Glide-In (OSG) with daily moth input and WRF-based input
```

```
universe = vanilla
log = ATM_WRF-NARR_${grid}_${begin_date}_default_${Process}.log
output = ATM_WRF-NARR_${grid}_${begin_date}_default_${Process}.out
error = ATM_WRF-NARR_${grid}_${begin_date}_default_${Process}.err
requirements = (OpSys == "LINUX") && (OpSysMajorVer >= 7)
transfer_input_files =
    miniconda.tar.gz,Circadian_calculations.py,Clock.py,Flier_class.py,Flier_grids.py,Flier_setup.py,Flier_summary.py,
    Geography.py,Interpolation.py,Map_class.py,Model_control.py,Model_initialization.py,Model_wrapup.py,Plots_gen.py,
    Radar_class.py,SBW_empirical.py,Simulation_specifications_WRF-NARR_${grid}_${begin_date}.py,Solar_calculations.py,
    Temporal_operations.py,WRFgrids_class.py,ATM_WRF-NARR_${grid}_${begin_date}_ready_moths.csv,
    WRF_out_reduced_${grid}_${begin_date}06-${end_date}12.tar.gz
executable = ATM_WRF-NARR.sh
arguments = ${grid} ${begin_date} ${end_date} ${Process}
request_cpus = 1
request_memory = 8GB
request_disk = 36GB
+WantFlocking = true
+WantGlideIn = true
queue ${nreps}
```

Python-based Atmospheric Transport Model (pyATM) for SBW moths

Stochastic simulations

- random subsets of available moths
- liftoff times vary with local conditions
- weather, esp. near the surface (BL)

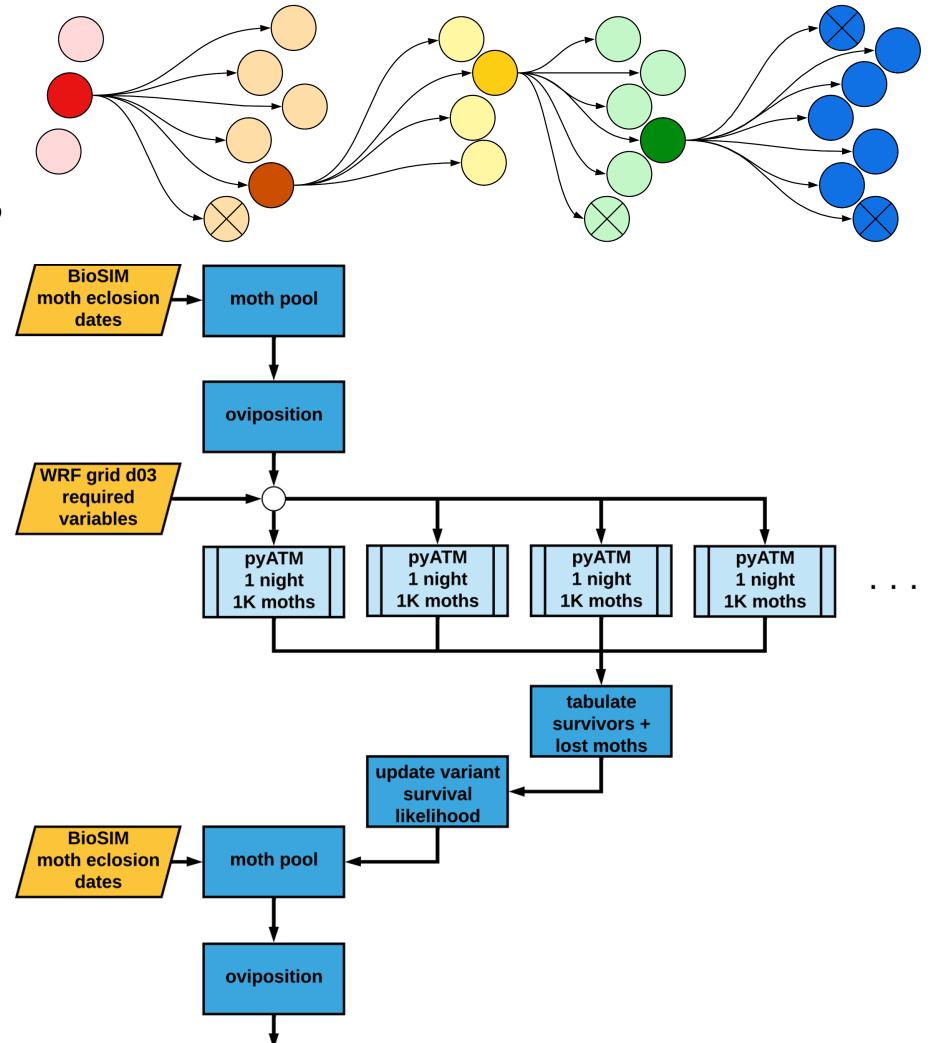
→ Monte Carlo ensemble of simulations

Sequential nightly simulations

- accounting variety of prior results
- compounded timelines + variants

→ Markov chain Monte Carlo process
with Bayesian likelihood updating

→ Probabilistic outcomes over the
mating + dispersal + oviposition
period



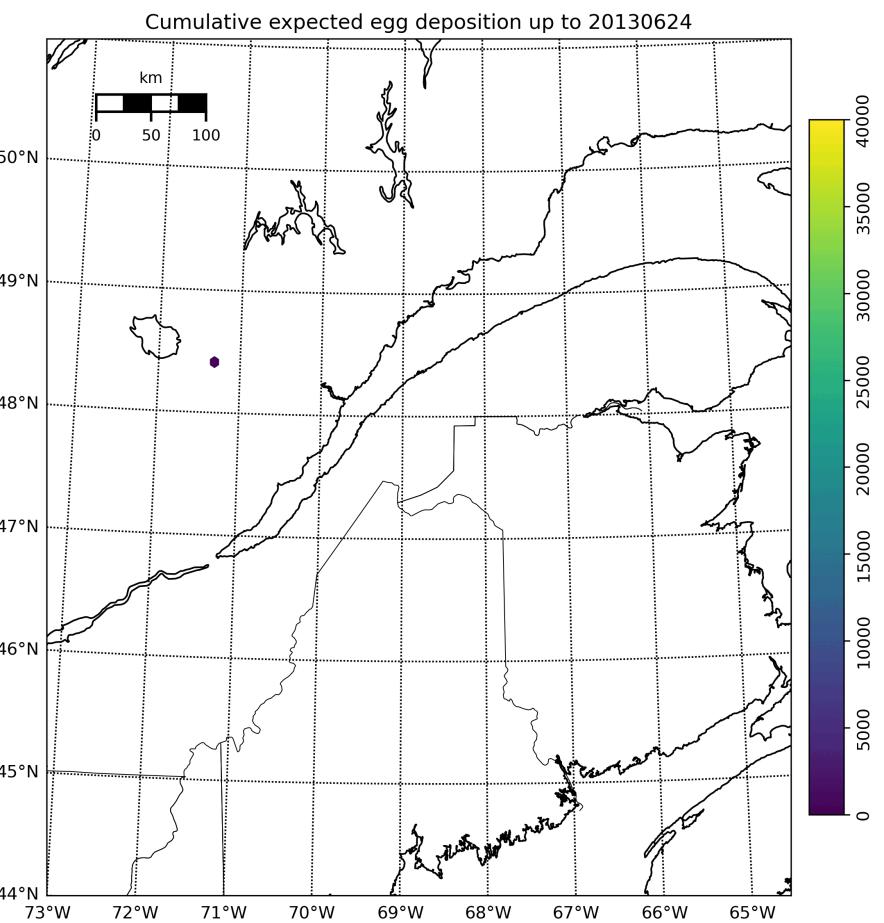
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→ daily changes in weight
and flight capability

Simulated 2013 daily cumulative egg deposition up to 9 July



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M immigration → increased poss.
of mating success

Simulated 2013 daily M landings + F eclosion up to 9 July

