Uncertainty, Stochastics & Sensitivity Analysis

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Types of Sensitivity Analyses

- Variables involved
  - One-way
  - Multi-way
- Type of component being varied
  - Parameter sensitivity analysis: Parameter values
  - Structural sensitivity analysis: Examine effects of model *structure* on results
- Type of variation
  - Single alternative values
  - Monte Carlo analyses: Draws from probability distributions (many types of variations)
- Frequency of variation
  - Static (parameter retains value all through simulation)
  - Ongoing change: Stochastic process
    - Accomplished via Monte-Carlo analyses
    - Key for DES & ABM

Model Uncertainty

- Here, we are frequently examining the impact of changing
  - Our assumptions about “how the system works”
  - Our decision of how to abstract the system behaviour
- Structural sensitivity analyses
  - Vary structure of model & see impact on
    - Results
    - Tradeoffs between choices
  - Frequently recalibrate the model in this process
- Here, we are considering uncertainty about how the current state is mapped to the next state

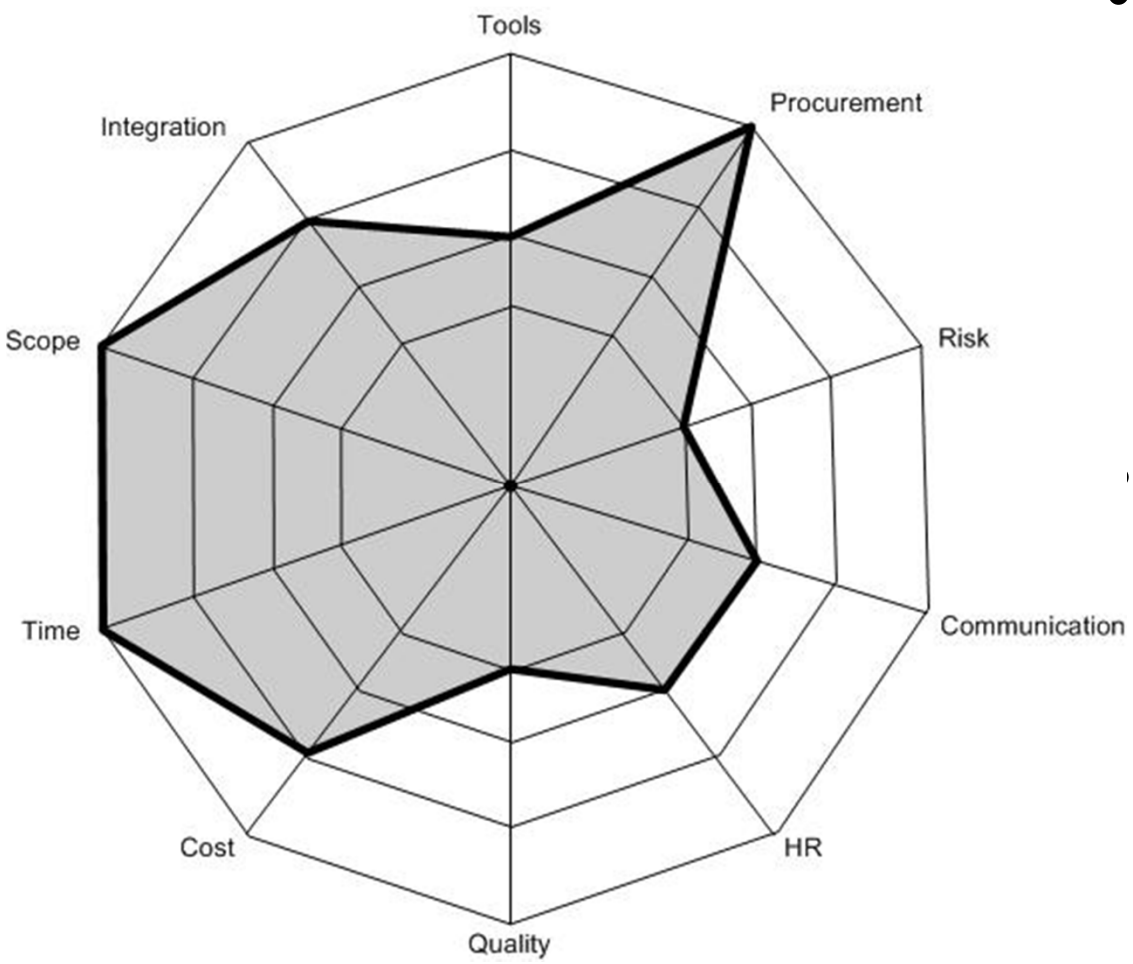
Predictor-Corrector Methods: Dealing with an Incomplete Model

- Some approaches (e.g. Kalman filter, Particle Filter) are motivated by awareness that models are incomplete
- Such approaches try to adjust model state estimates on an ongoing basis,
  - Given uncertainty about model predictions
  - New observations
- Assumption here is that the error in the model is defined by some probability distribution

Static Uncertainty Sensitivity Analyses

- In variation, one can seek to investigate different
  - Assumptions
  - Policies
- Same relative or absolute uncertainty in different parameters may have hugely different effect on outcomes or decisions
- Help identify parameters/initial states that strongly affect
  - Key model results
  - Choice between policies
- We place more emphasis in parameter estimation & interventions into parameters exhibiting high sensitivity

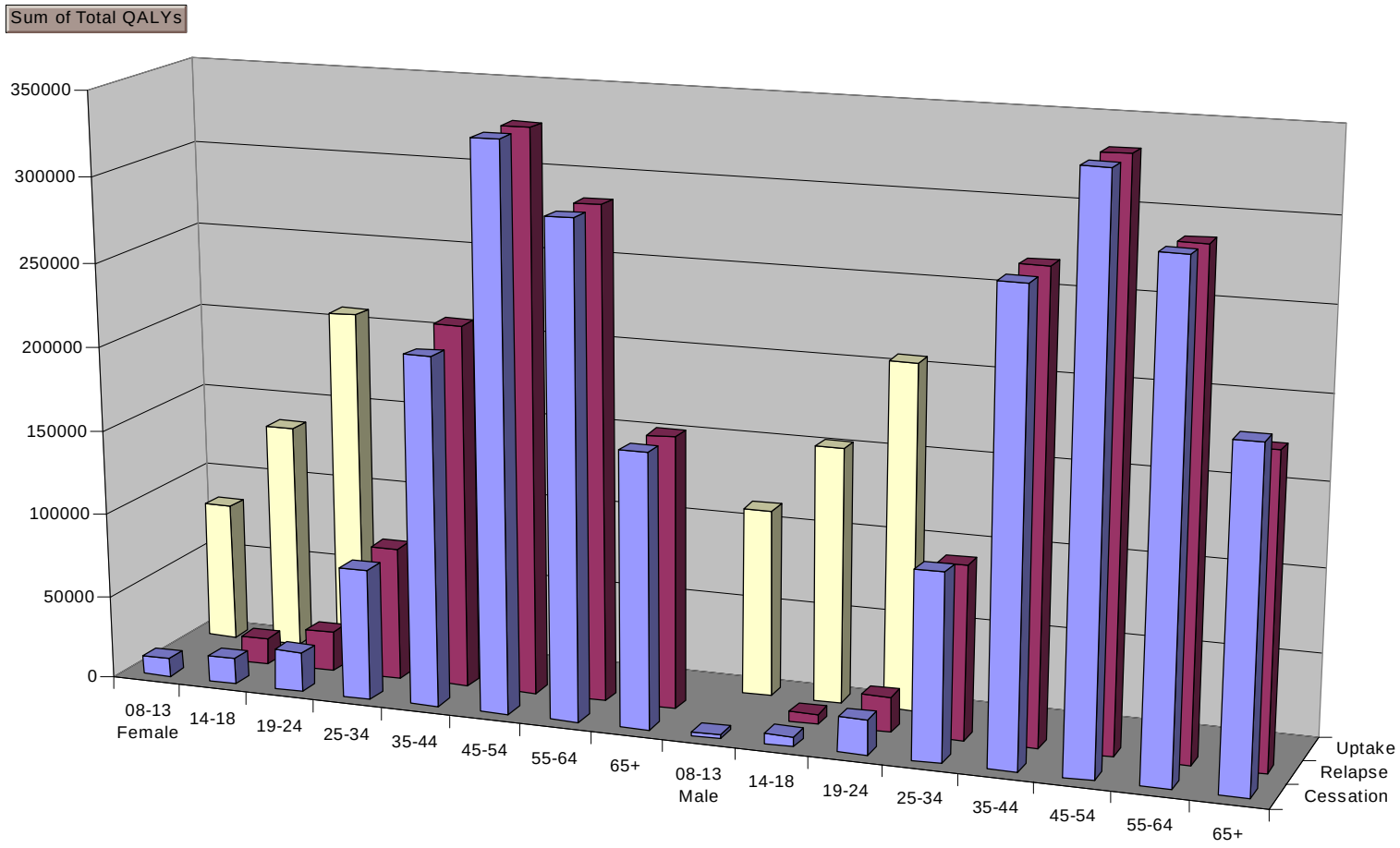
Spider Diagram



- Each axis represents a % change in a particular parameter
  - This proportional change is identical for the different parameters
- The distance assumed by the curve along that axis represents the magnitude of response to that change
  - Note that these sensitivities will depend on the state of system!

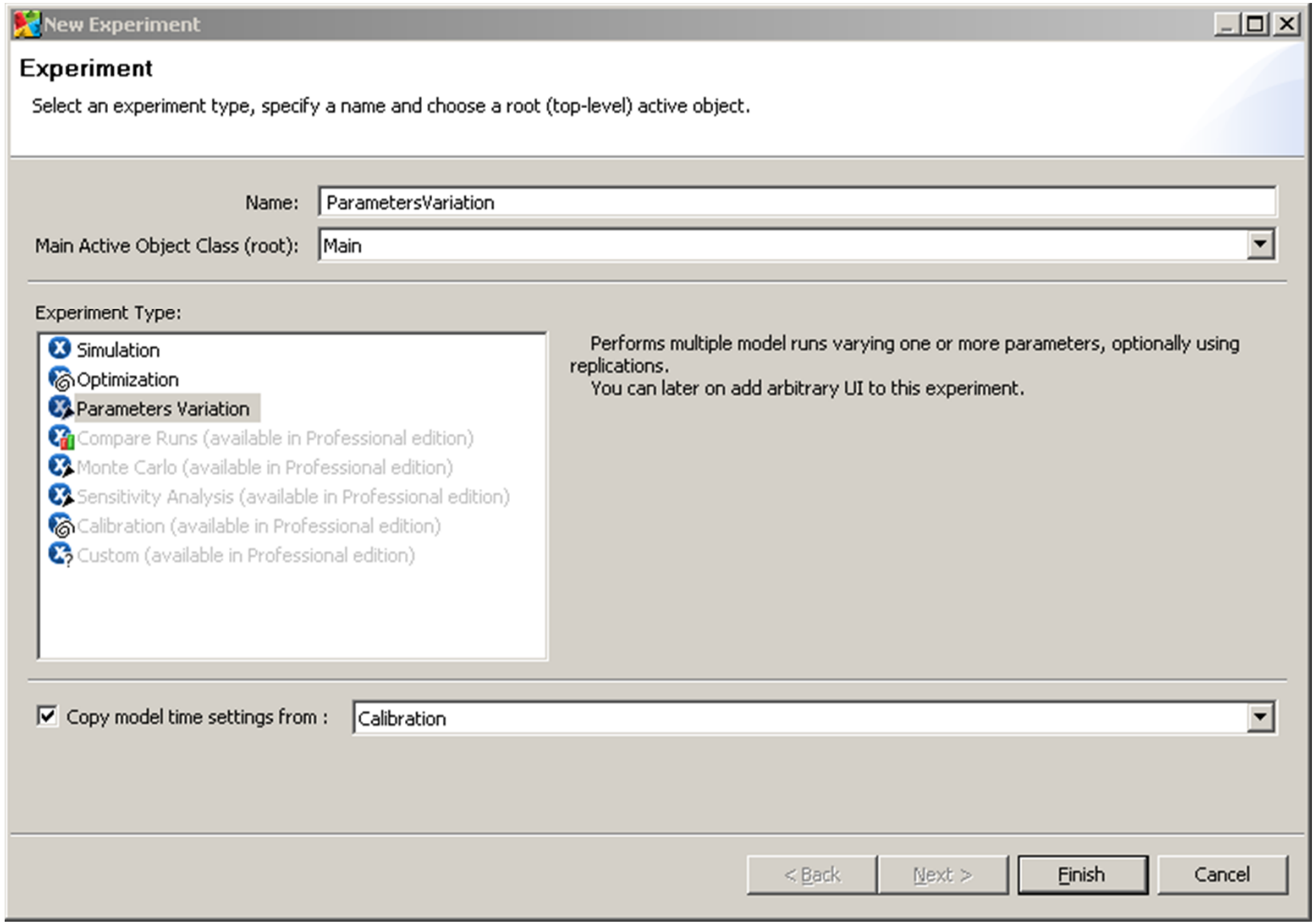
http://www.niwotridge.com/images/BLOGImages/SpiderDiagram.jpg

Systematic Examination of Policies

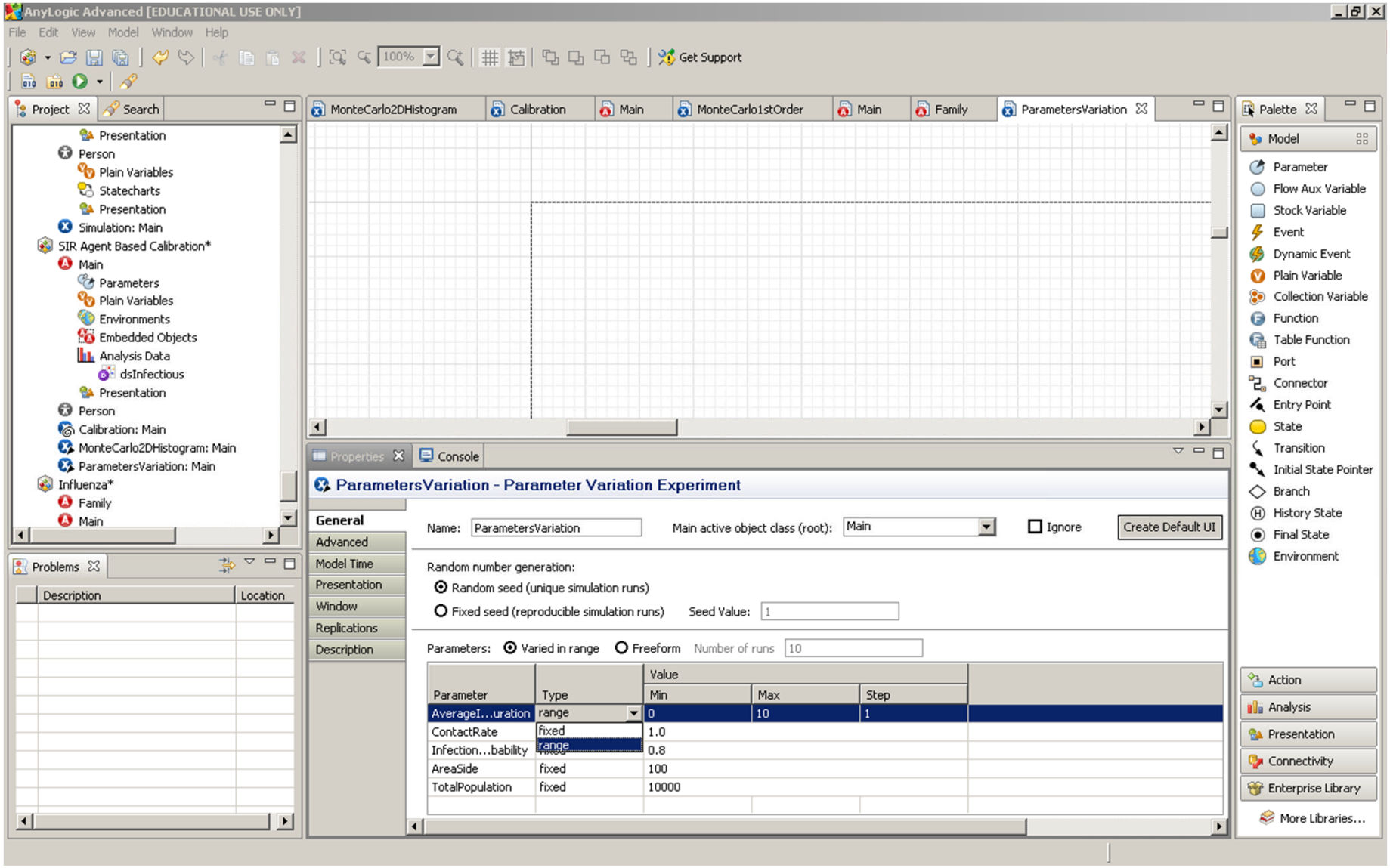


Tengs, Osgood, Lin

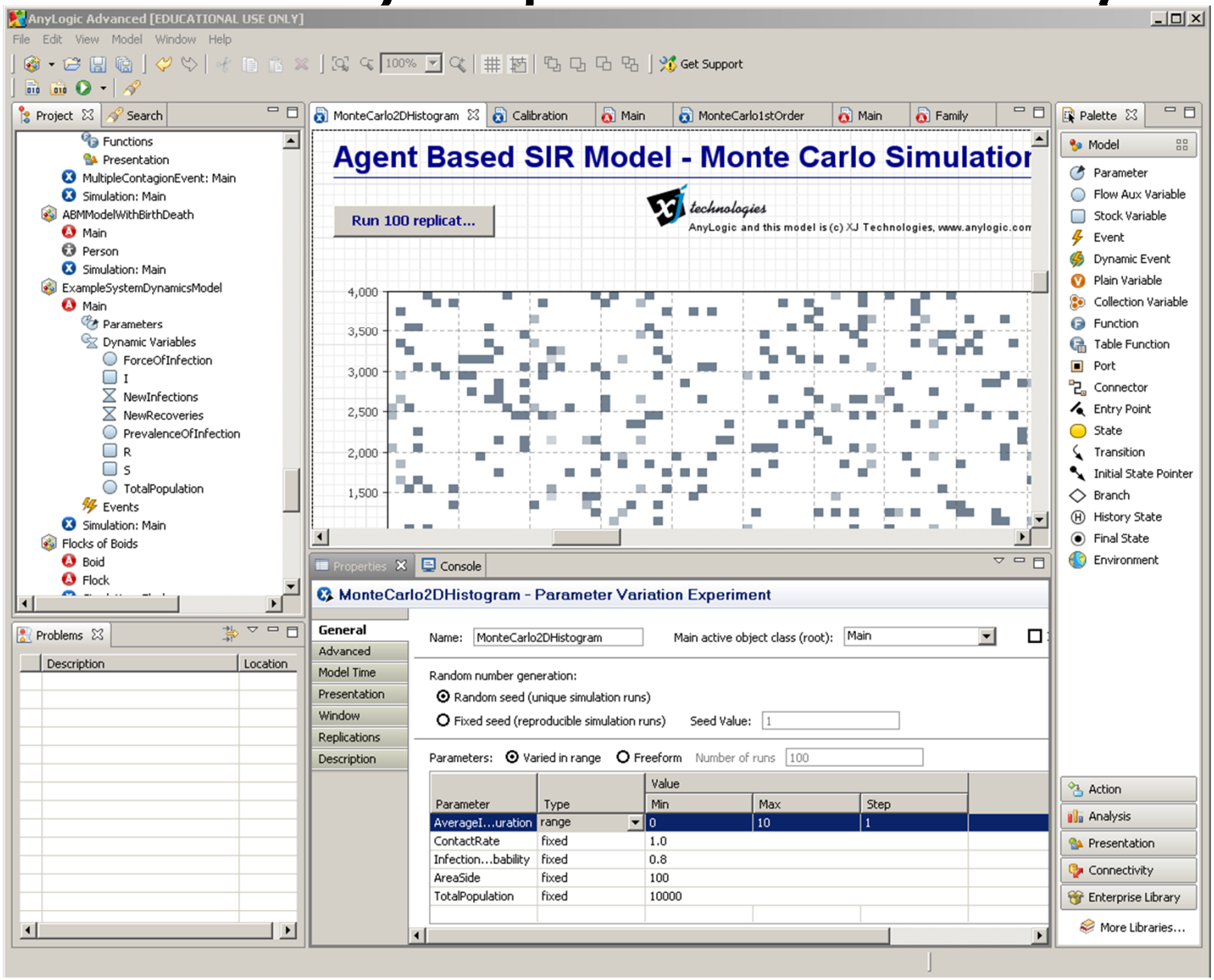
Add New “Parameters Variation” Experiment



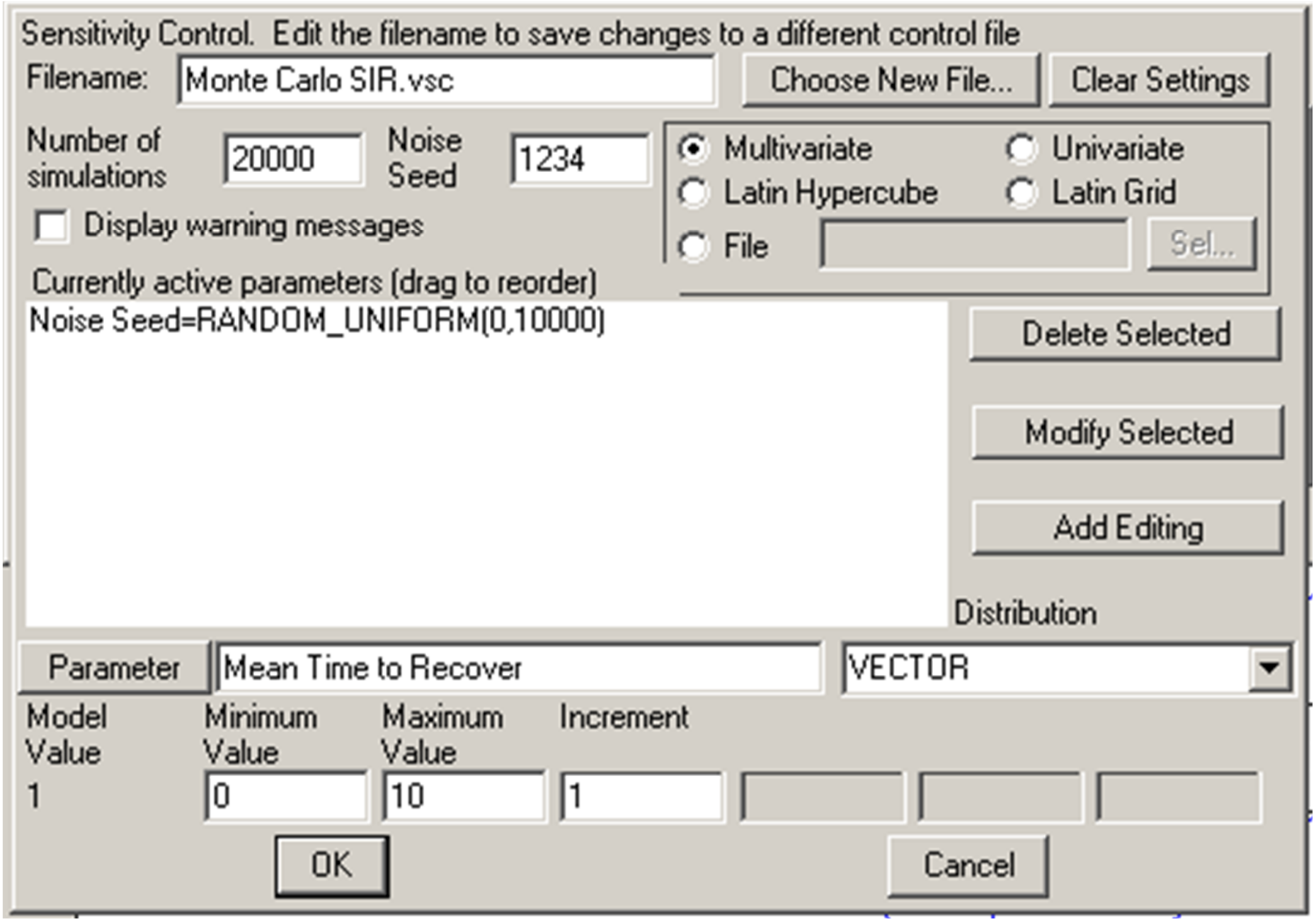
Setting Ranges for Parameter Variation



Sensitivity Exploration in AnyLogic



Sensitivity Analyses in Vensim



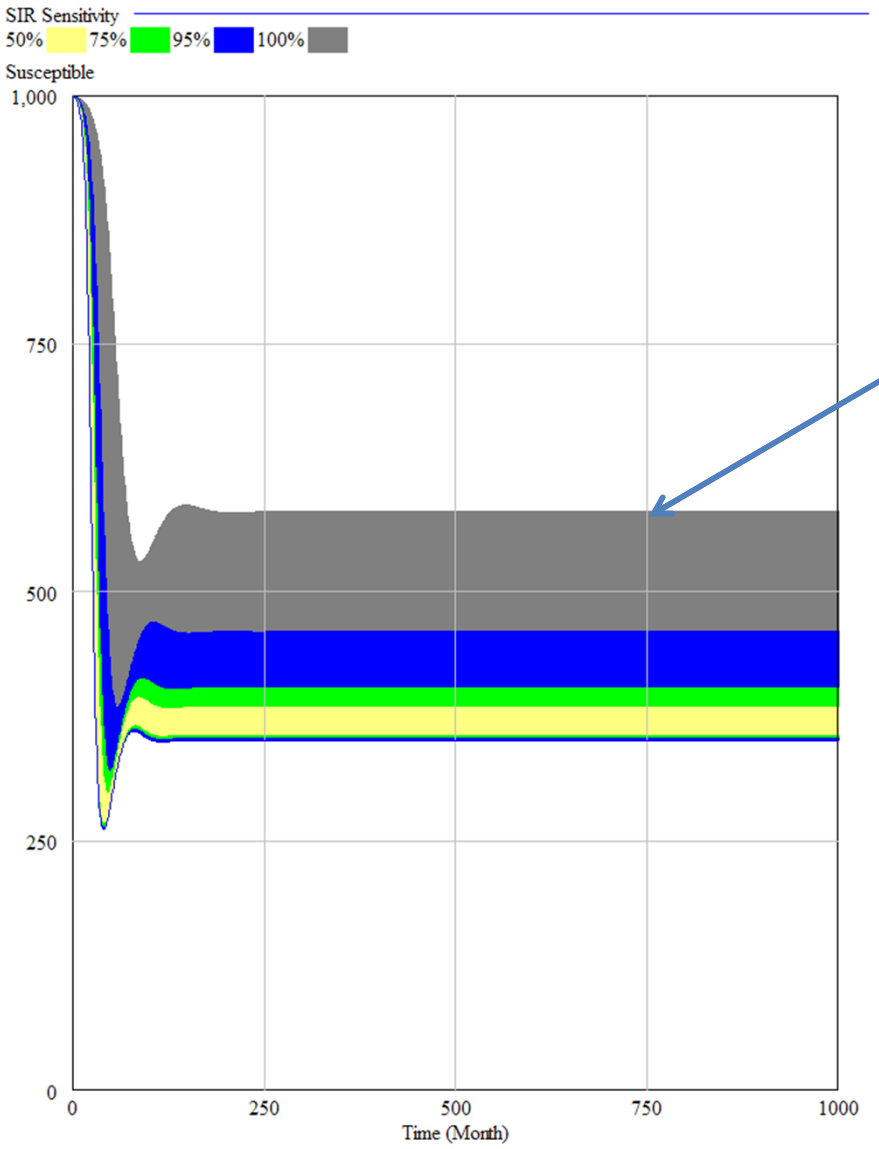
Sensitivity in Initial States

- Frequently we don't know the exact state of the system at a certain point in time
- A very useful type of sensitivity analysis is to vary the initial model state
- In Vensim, this can be accomplished by
  - Indicating a parameter name within the "initial value" area for a stock
  - Varying the parameter value
- In an agent-based model, state has far larger dimensionality
  - Can modify different numbers of people with characteristic, location of people with characteristic, etc.

Imposing a Probability Distribution Monte Carlo Analysis

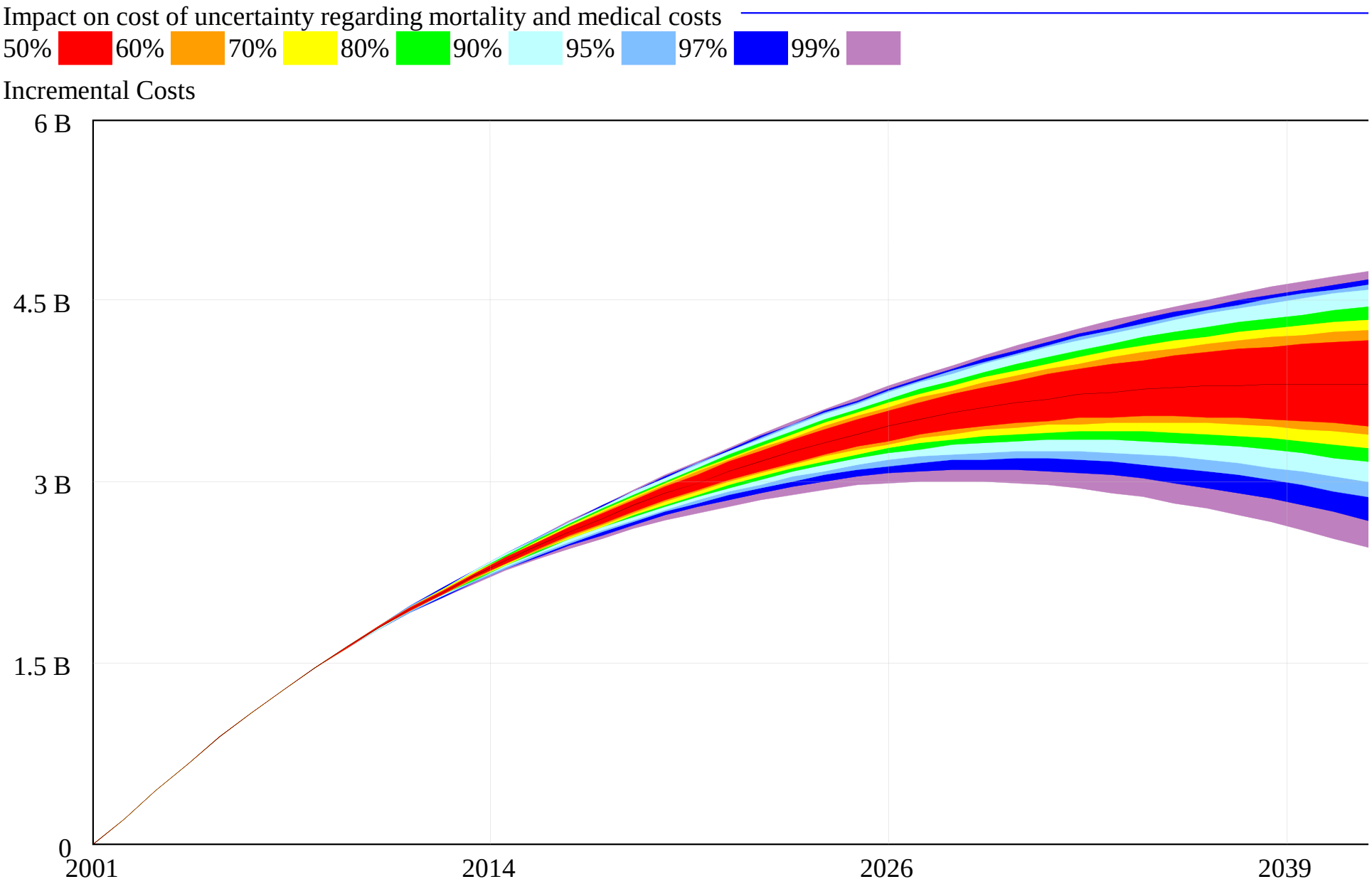
- We feed in probability distributions to reflect our uncertainty about one or more parameters
- The model is run many, many times (realizations)
  - For each realization, the model uses a different draw from those probability distribution
- What emerges is resulting probability distribution for model outputs

Example Resulting Distribution



Empirical Fractiles

Static Uncertainty



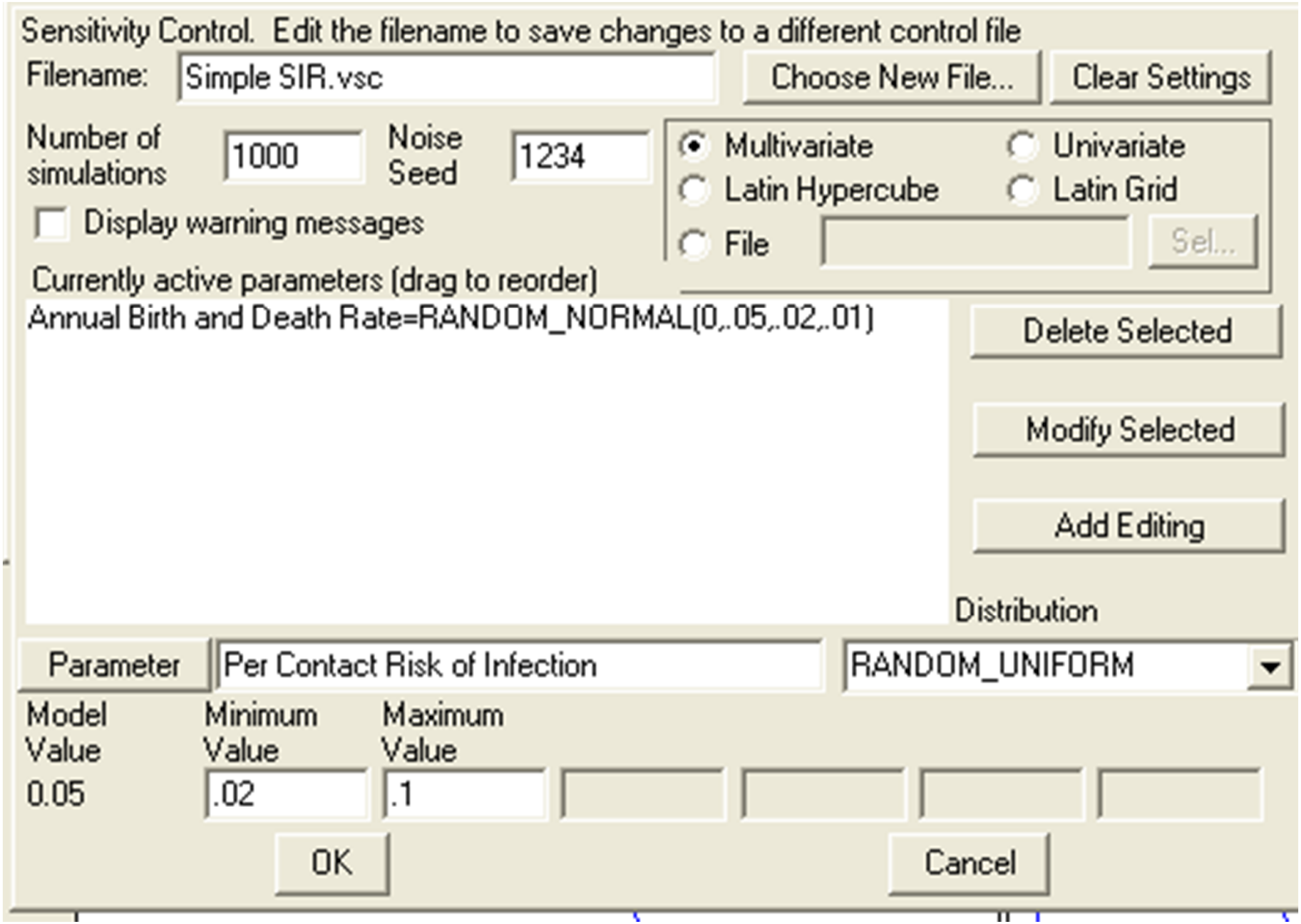
Multi-Way Sensitivity Analyses

- When examining the results of changing multiple variables, need to consider how multiple variables vary together
- If this covariation reflects dependence on some underlying factor, may be able to simulate uncertainty in underlying factor

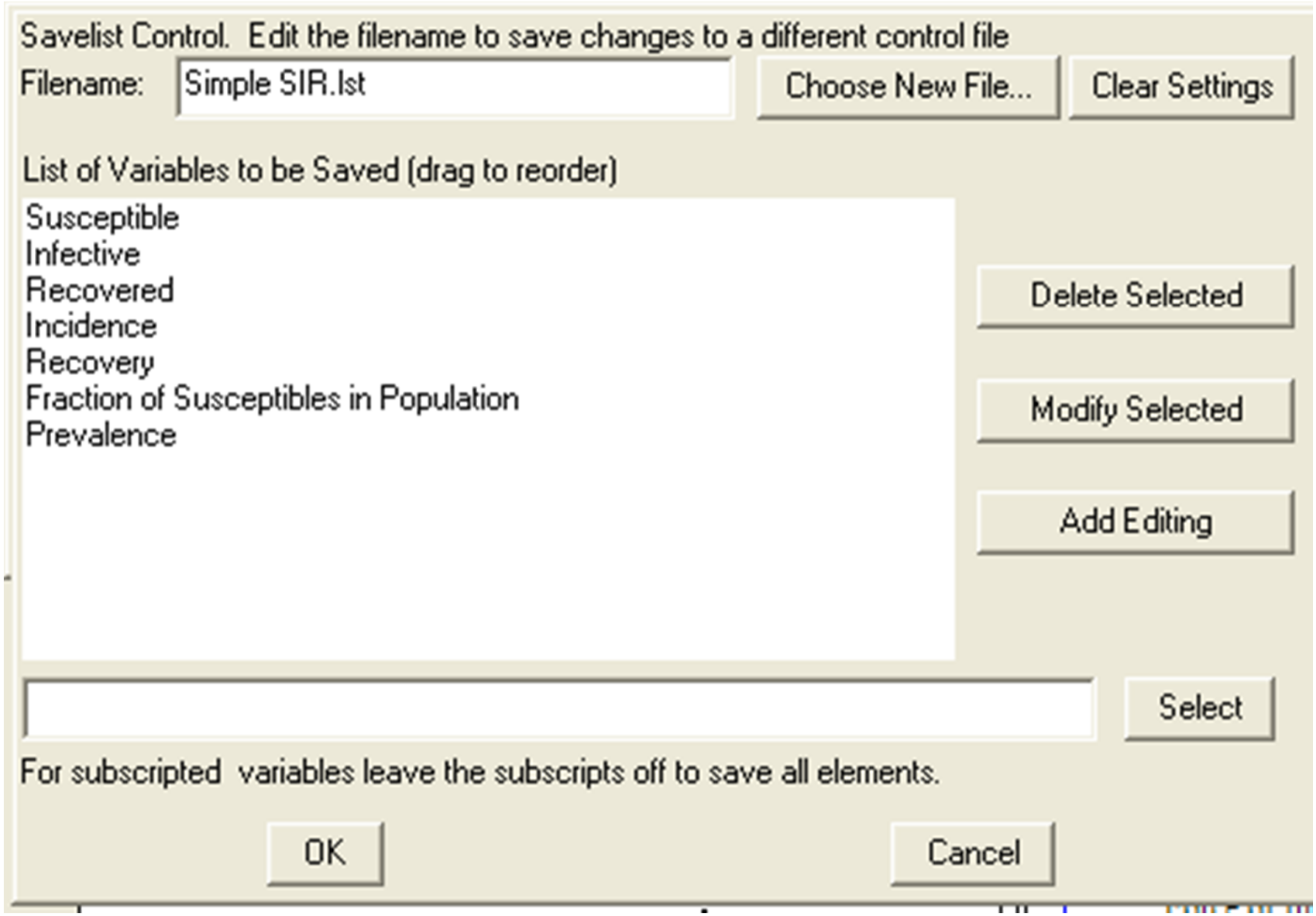
Performing Monte Carlo Sensitivity Analyses in Vensim

- Need to specify three things
  - The parameters to vary
  - How to vary those parameters
  - Which model variables to save away

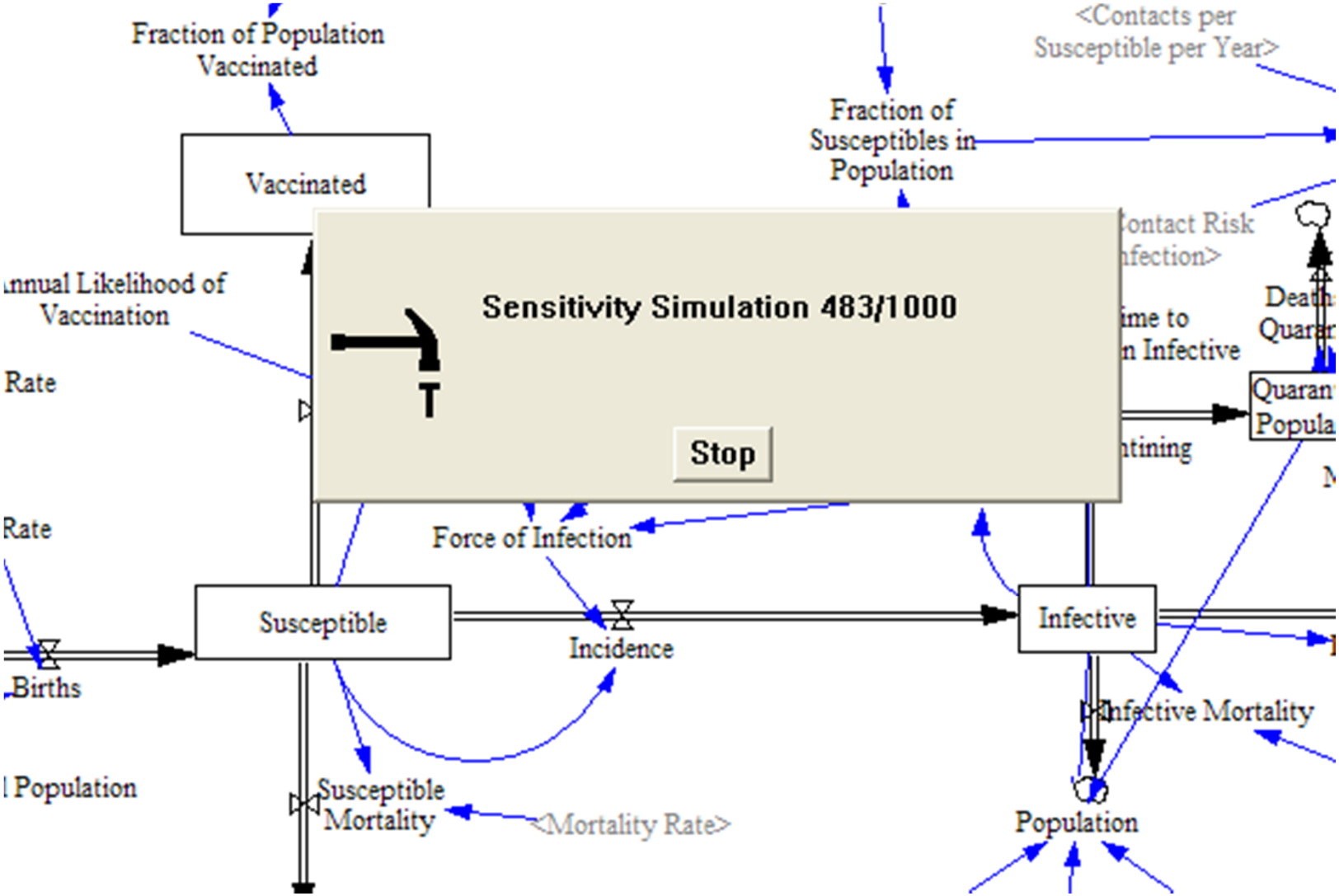
How & What Parameters to Vary



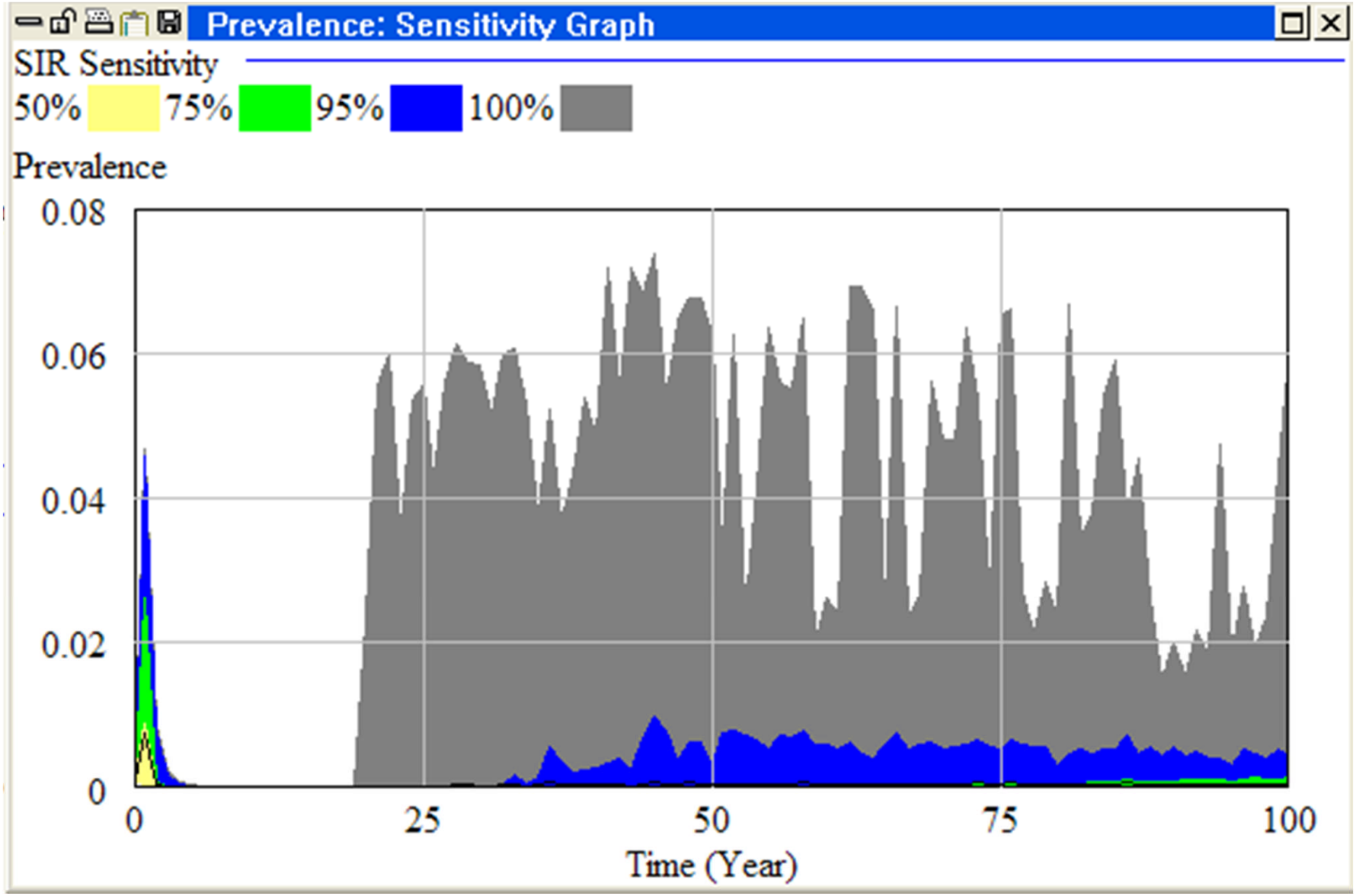
Model Values to Save Away



Monte Carlo Analyses

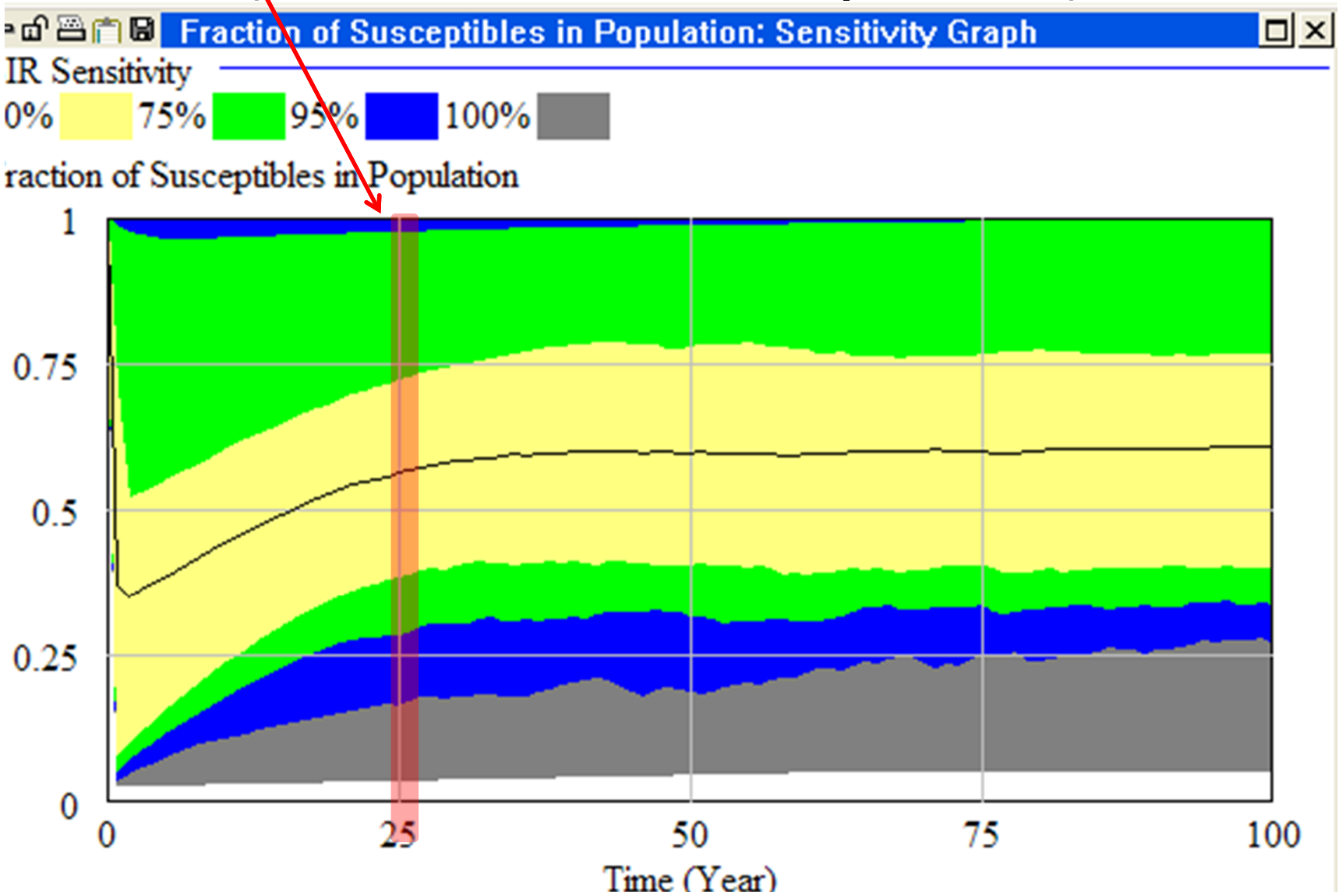


Sensitivity Results (Prevalence)



An observation at this point in time would produce a histogram (approximating a distribution) for fraction of susceptibles

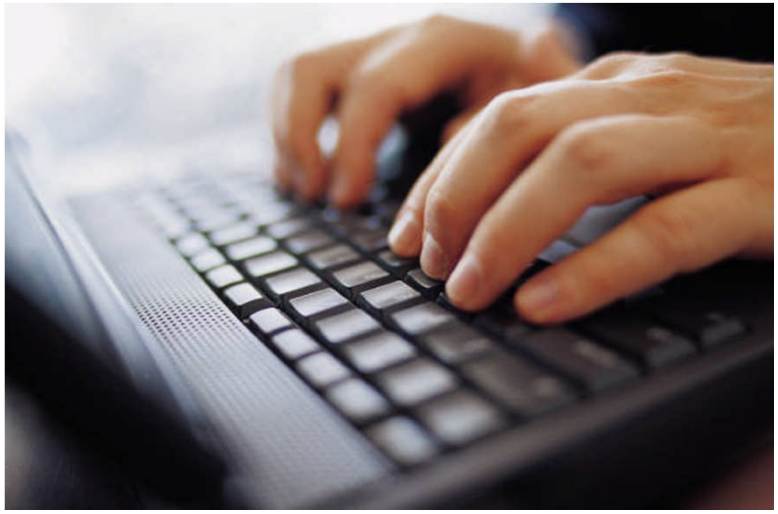
Sensitivity Results (Fraction of Susceptibles)



Monte Carlo Analyses in AnyLogic

- When running Monte Carlo analysis, we'd like to summarize the results of multiple runs
- One option would be to display each trajectory over time; downside: quickly gets messy
- AnyLogic's solution
  - Accumulate data regarding how many trajectories fall within given areas of value for a given interval of time using a "Histogram2D Data"
  - Display the Histogram2D Chart

Hands on Model Use Ahead

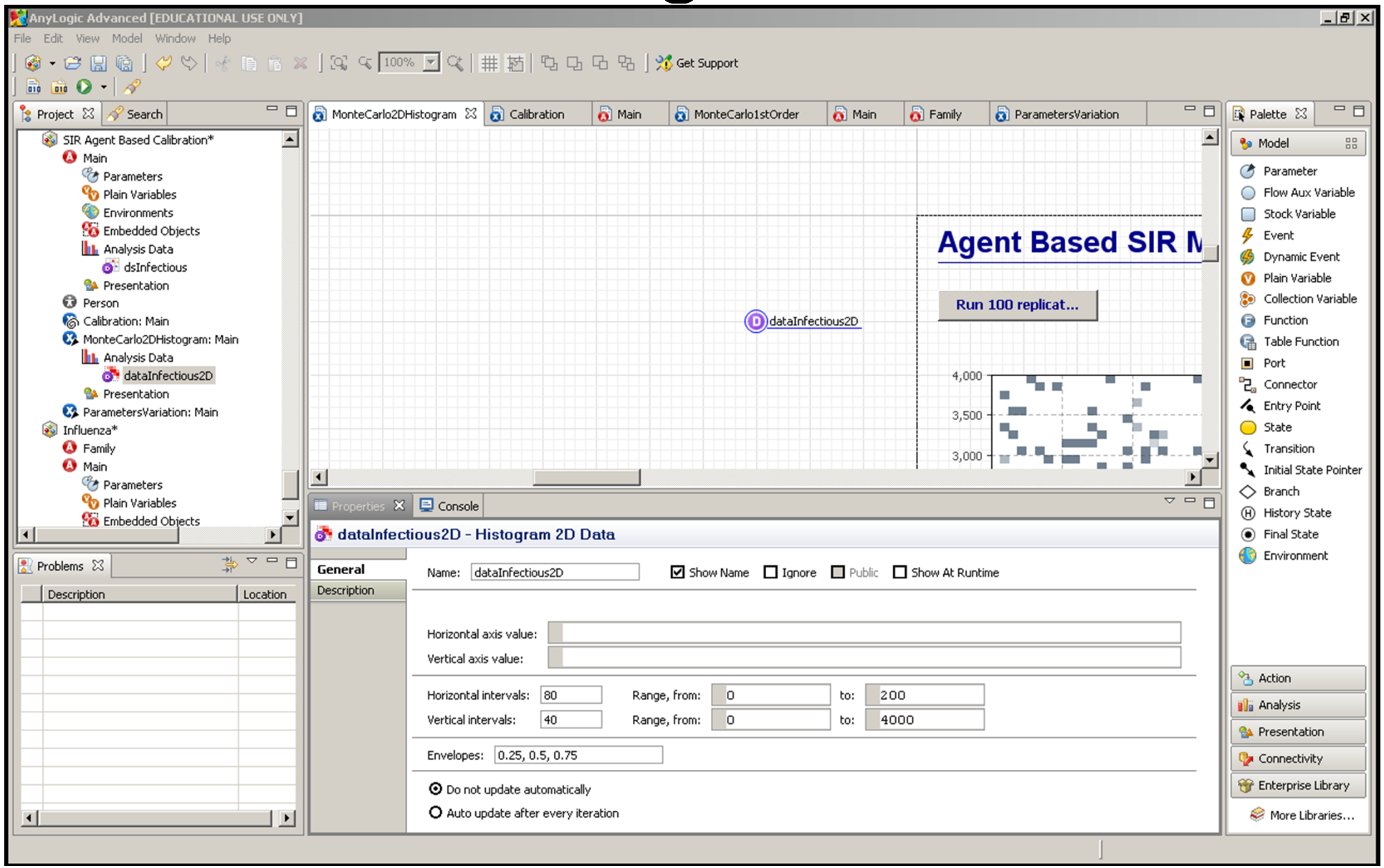


Load Sample Model:

SIR Agent Based Calibration

(Via “Sample Models” under “Help” Menu)

2D Histogram Data



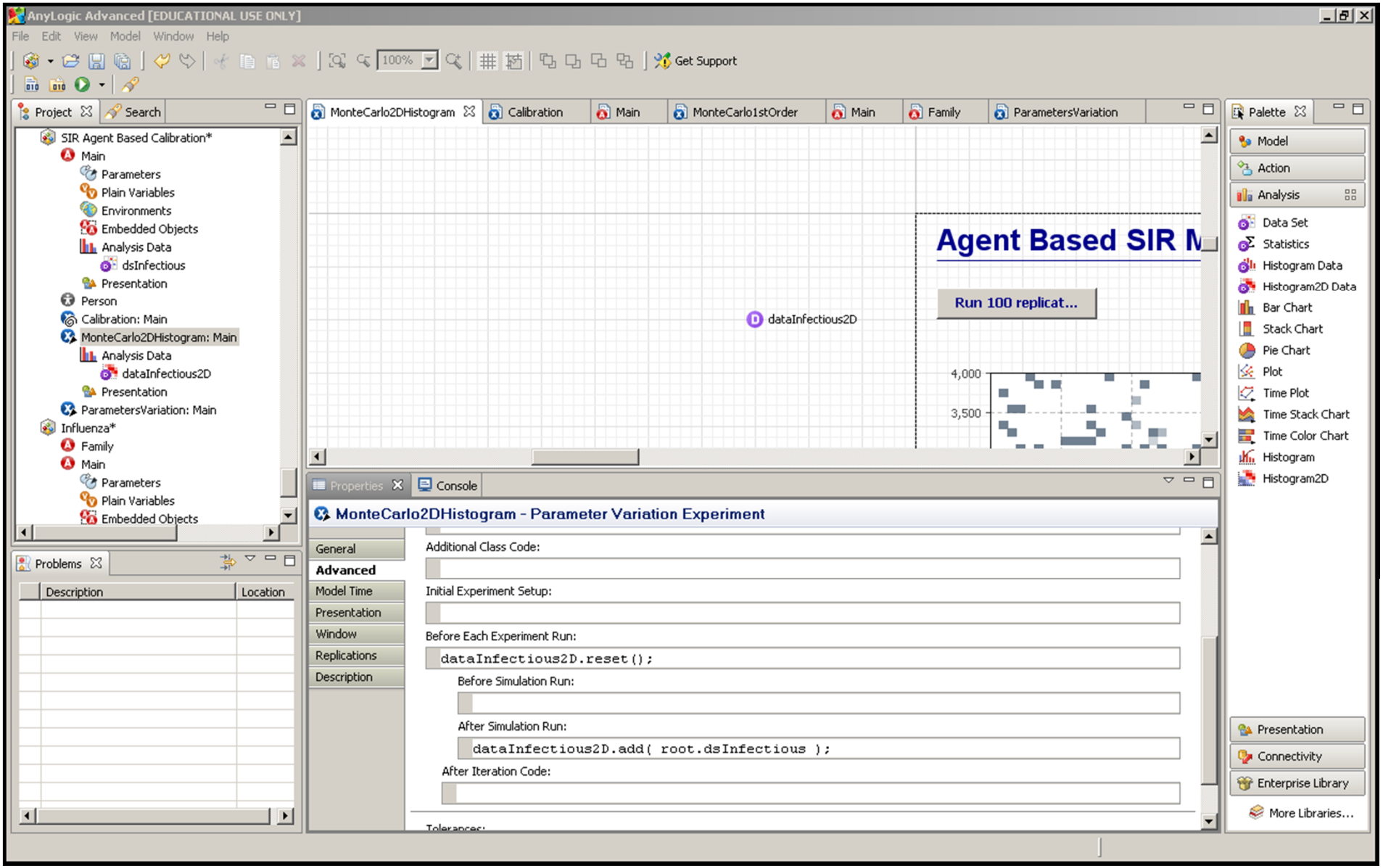
Important Distinction (Declining Order of Aggregation)

- Experiment
  - Collection of simulation
- Simulation
  - Collection of replications that can yield findings across set of replications (e.g. mean value)
- Replication
  - One run of the model

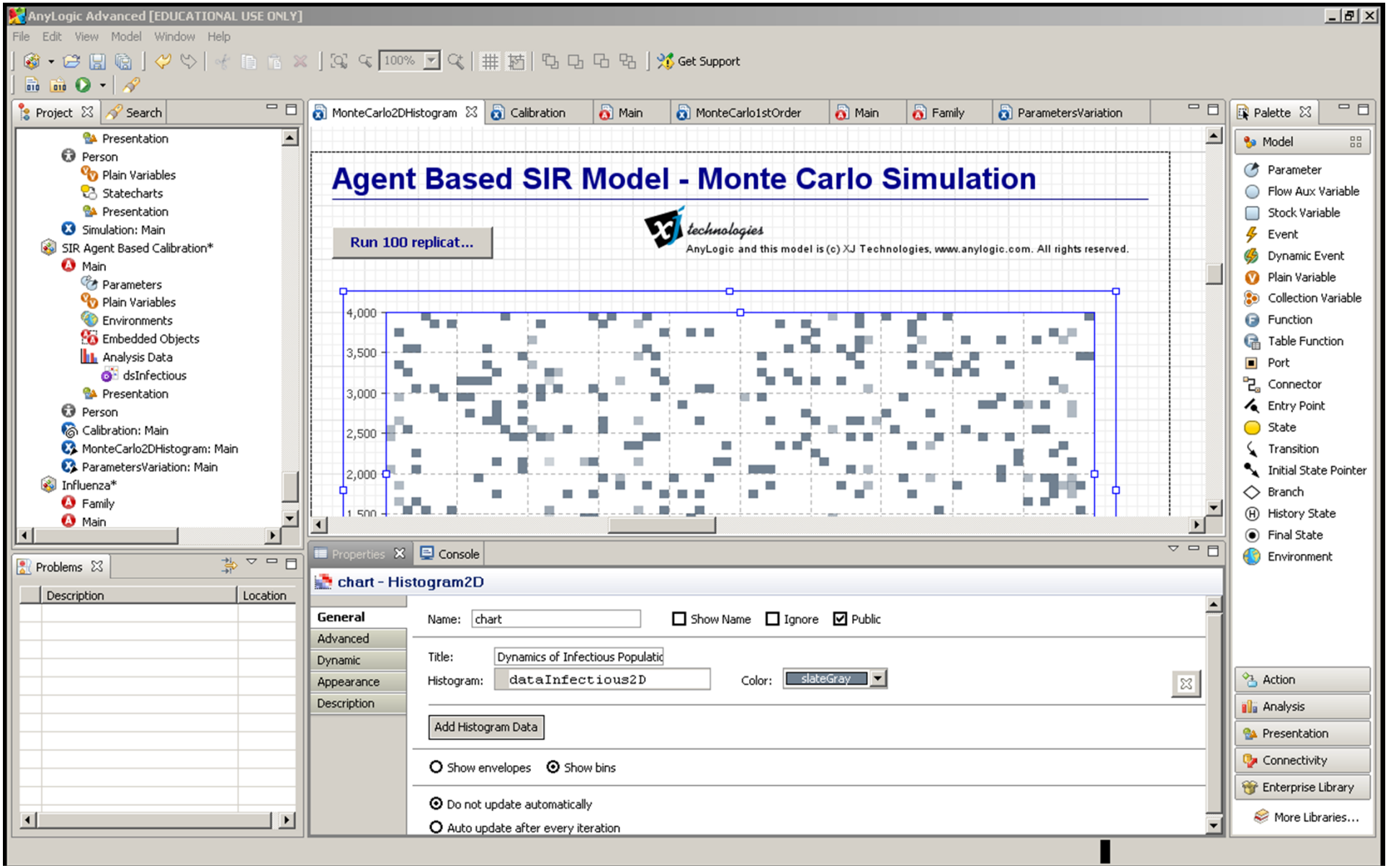
Flexibility Typically Ignored

- In most AnyLogic models, an Experiment is composed of a single Simulation, which is composed of a single Replication
- In most AnyLogic models which run "ensembles" of realizations, a simulation is composed of only a single realization

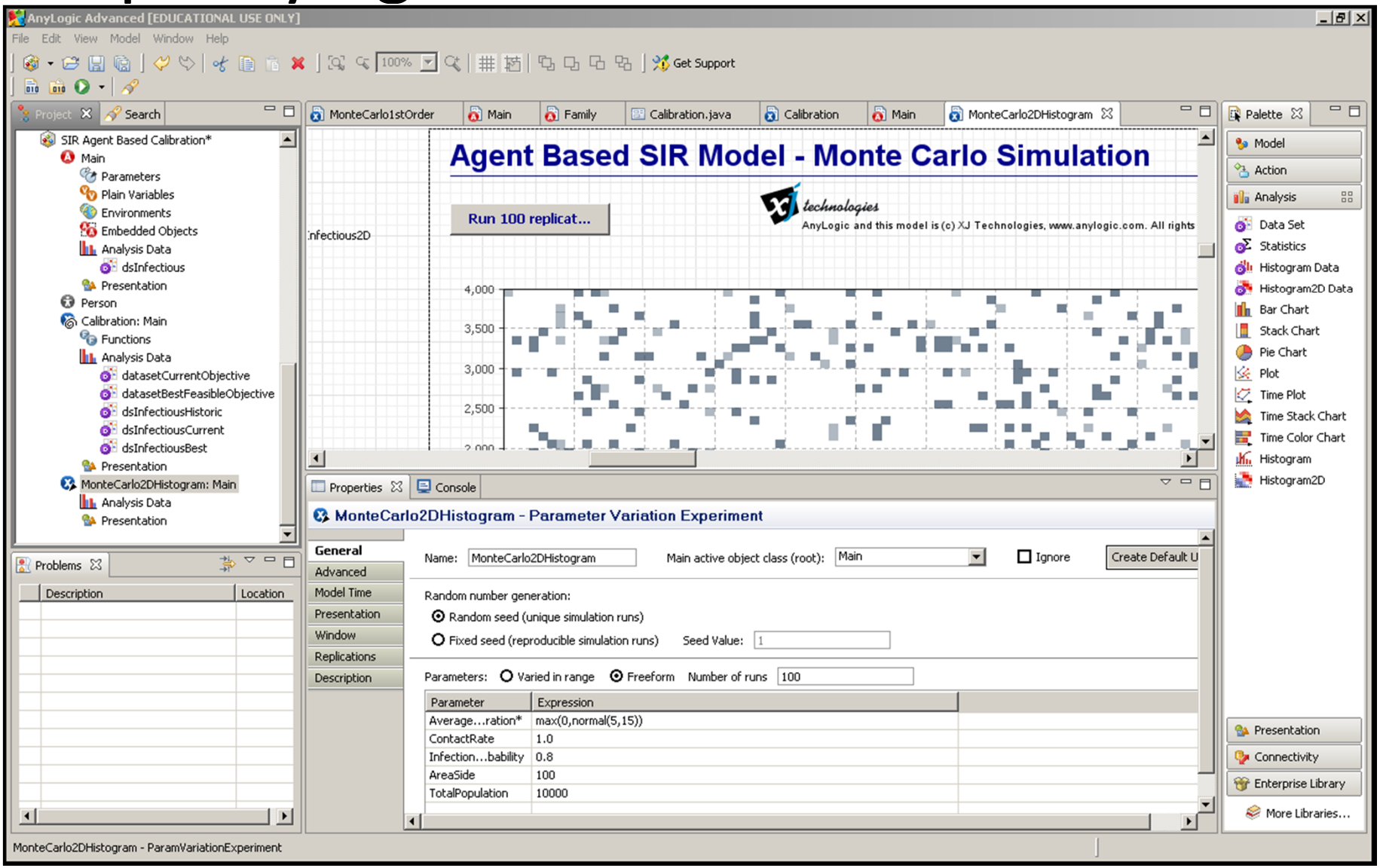
Accumulating the Histogram2D dataset from other datasets



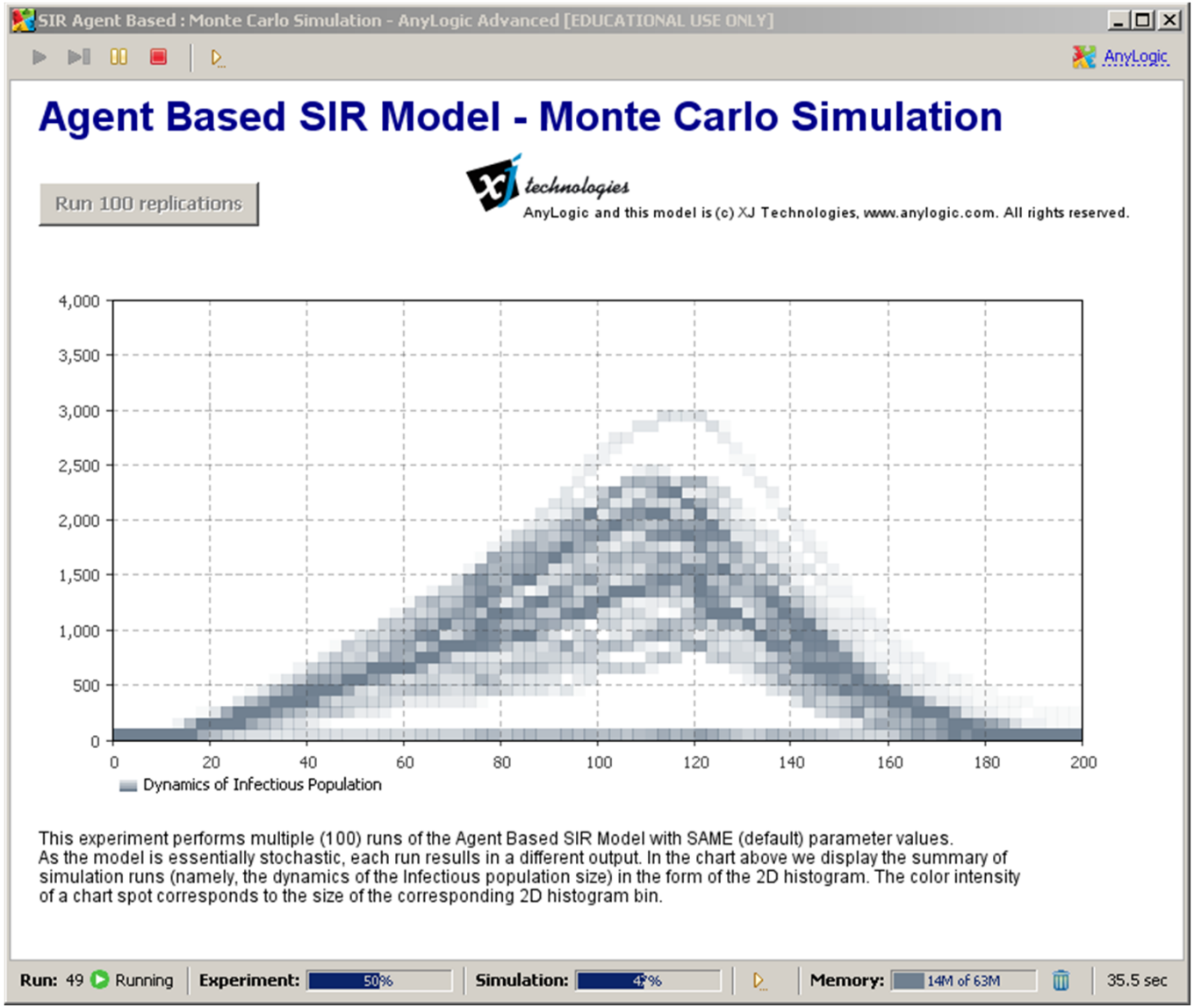
Monte Carlo Sensitivity Analyses in AnyLogic



Monte Carlo Analyses in AnyLogic: Specifying Distributions for Parameters



Monte Carlo Output
After All Runs



Monte Carlo Output
After All Runs