VENTANA systems, inc.

Vensim® Software

Linking systems thinking to powerful dynamic models

Calibration with Vensim

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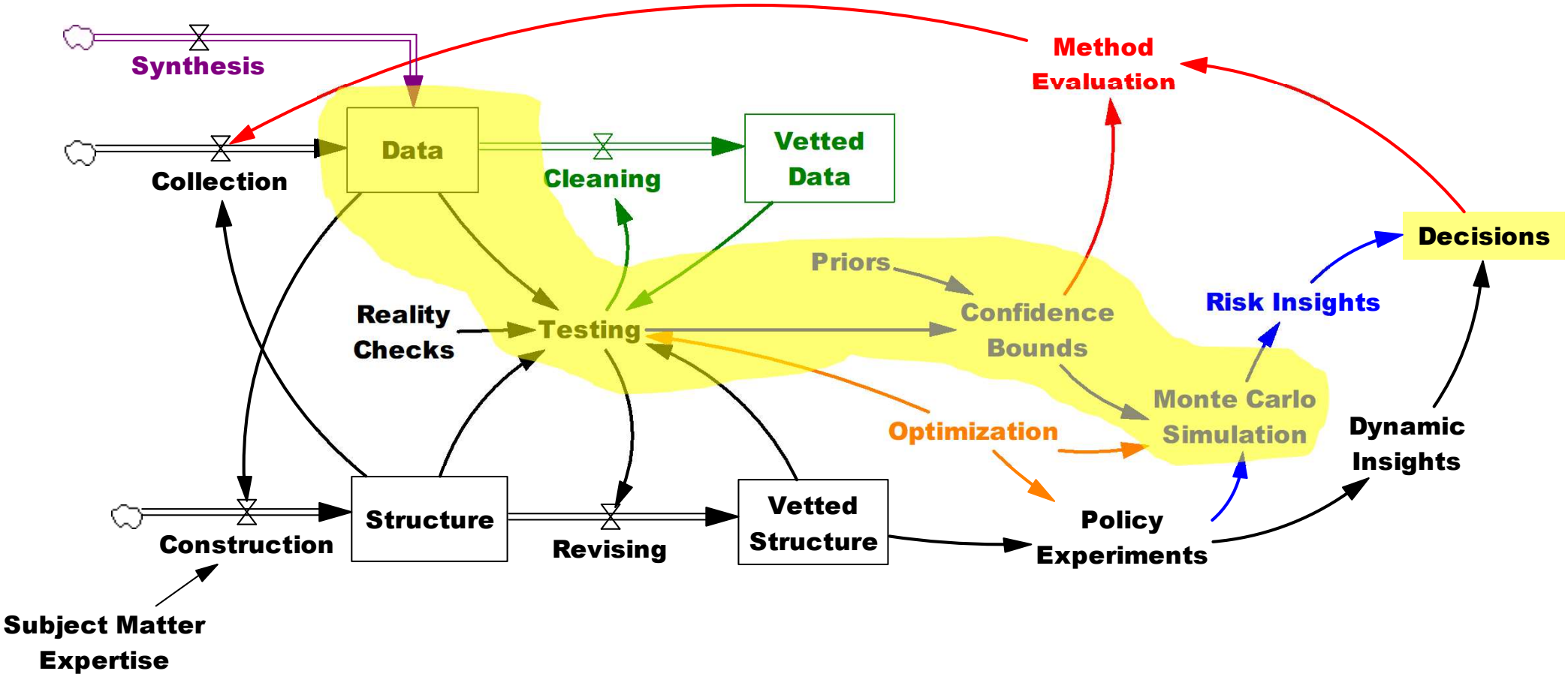
2022

Agenda

- **Introduction**
- **Synthetic data**
- **Naïve calibration**
- **Maximum likelihood**
- **Markov Chain Monte Carlo (MCMC)**
- **Kalman filtering (briefly)**

The Big Picture



Overview

- **Lotka-Volterra Predator-Prey model**

- **Generate synthetic data, by adding:**
  - Measurement error to the stocks of elk and wolves
  - Driving noise to the flows of births and mortality

- **Estimate parameters of the model from the data, by various methods**
  - Optionally, use mismatched structure (2nd order data-generating model, first order estimated model)

Caveats

In order to get done, we're approaching this problem a bit fast and loose. Be aware:

- **There is structural uncertainty as well as parameter uncertainty**
- **Statistics deserve deeper thought**
  - Weights
  - Covariance
  - Autocorrelation
  - Distributional assumptions
  - Measurement error & driving noise (Kalman filter)
- **We should be testing for multiple optima with multistart calibration runs**
- **Sample sizes for sensitivity and MCMC may be too small**

Other Pitfalls

- State dependent noise
- Sample size
- Data quality
- Autocorrelated errors
- Error covariance
- Measurement error
- State estimation
- Endogeneity

Calibration

- **Purposes**
  - Make better predictions or measurements
  - Reject models that can't replicate data (potentially a weak test of quality)
  - Learn about the model
  - Learn about the data
  - Provide face validity for reviewers
- **Closeness Measures**
  - Sum of squared errors & R^2
  - Mean Absolute Deviation
  - Mean Absolute Percent Error
  - Log Likelihood
- **Process**
  - Assume the model structure is right
    - If possible, test alternatives!
  - Simulate the model
    - Measure the closeness
    - Adjust the constants in the model
    - Iterate to improve
  - After convergence, evaluate the fit
    - Decide if the model needs revision
    - Investigate puzzles in the data

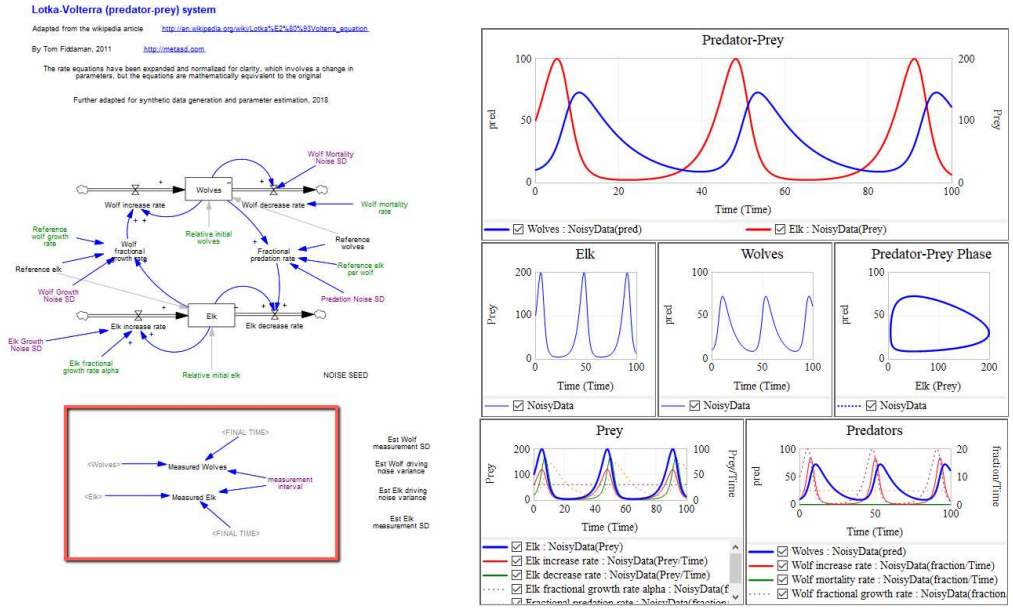
Model Tour

Synthetic Data Generation

- **Open the data-generating model**

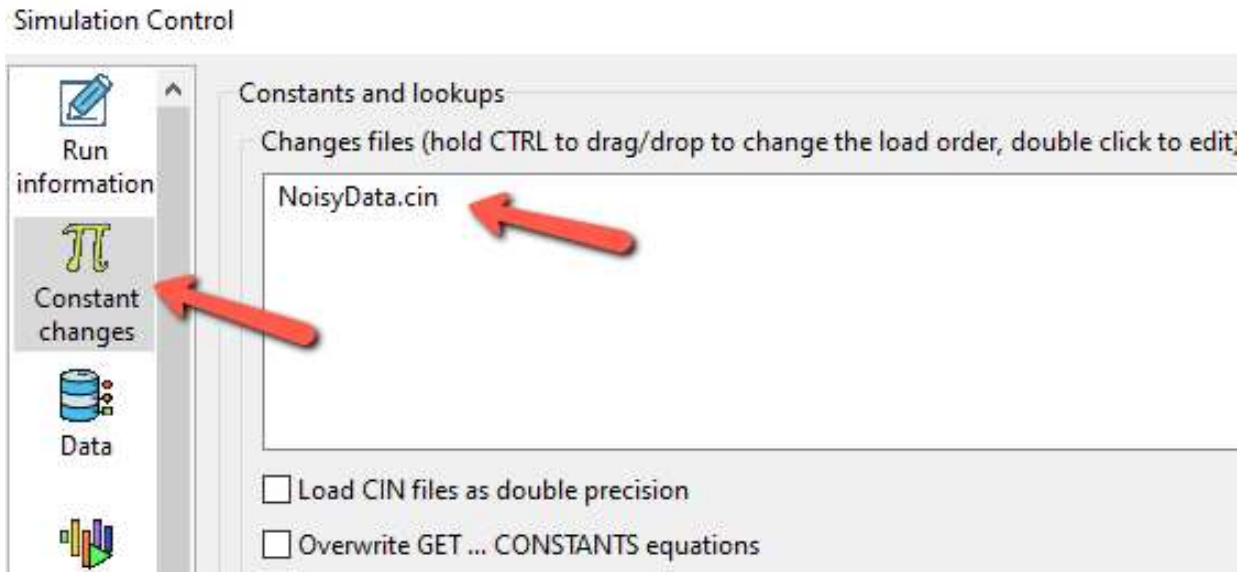
Name	Status	Date modified	Type	Size
ElkWolves - estimate - kalman	✓	7/23/2021 2:02 PM	File folder	
ElkWolves - estimate - mcmc	✓	7/23/2021 2:06 PM	File folder	
ElkWolves - estimate - naive	✓	7/23/2021 2:02 PM	File folder	
ElkWolves - estimate - start	✓	7/23/2021 2:03 PM	File folder	
ElkWolves - estimate - weighted	✓	7/23/2021 2:03 PM	File folder	
ElkWolves - generate	↻	7/28/2021 7:17 PM	File folder	
ElkWolves - generate - 2o	✓	7/23/2021 2:03 PM	File folder	

- **Run, and take a look at the "measured" variables (red box)**

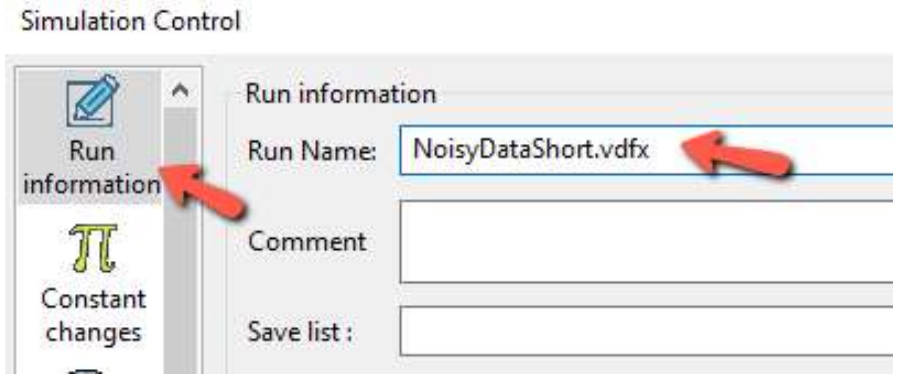


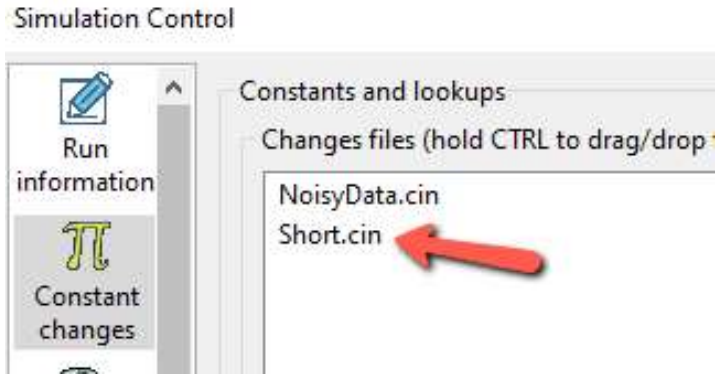
Synthetic Data Generation

- **Load "NoisyData.cin" and run the model**



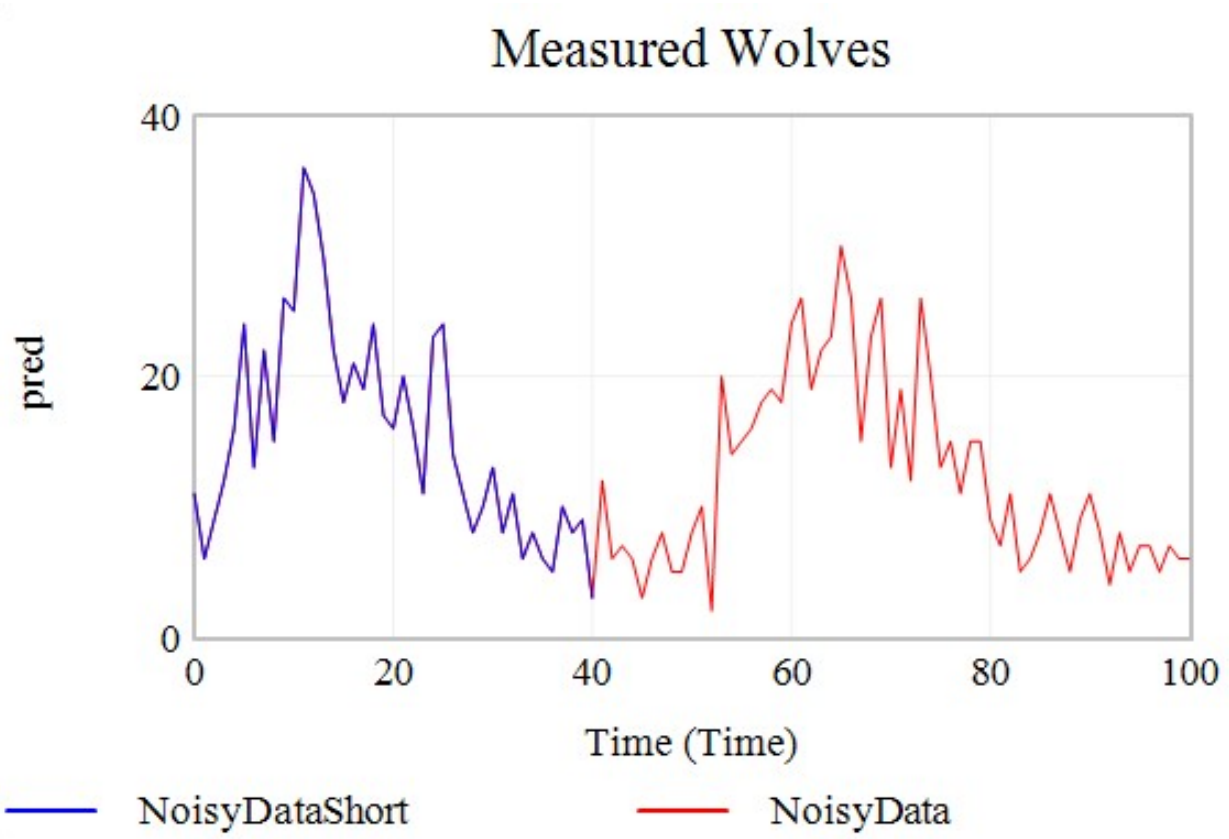
- **Run again, but add "short" to the run name, and load "Short.cin" to set FINAL TIME=40**





Synthetic Data Generation

- **This gives you 2 datasets**
  - Short time series, used to estimate parameters; export and re-import using a savelist to restrict the information to the measured state only
  - Long time series, for comparison of later estimates with "truth"



Naïve Calibration

- **Create a simple metric describing the distance of the model from the data**
- **Minimize the distance**

Mechanics – What We Need

- **Data**
  - A .vdf file, or
  - Equations with GET DIRECT DATA, GET XLS DATA
  - ODBC

- **A Payoff specifying what data series to match, and how to weight each one**

- **An Optimization Control file specifying**
  - which parameters to vary, and
  - what methods to use

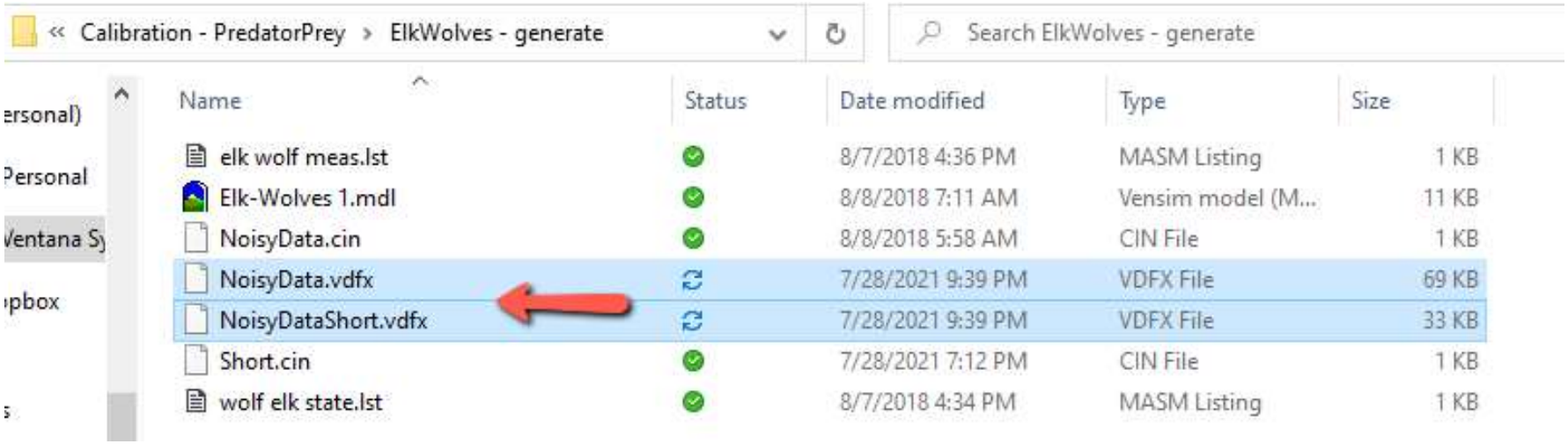
- **The optimizer then hill climbs to find the parameters that minimize the error between model and data**

Naïve Calibration
What do we get?

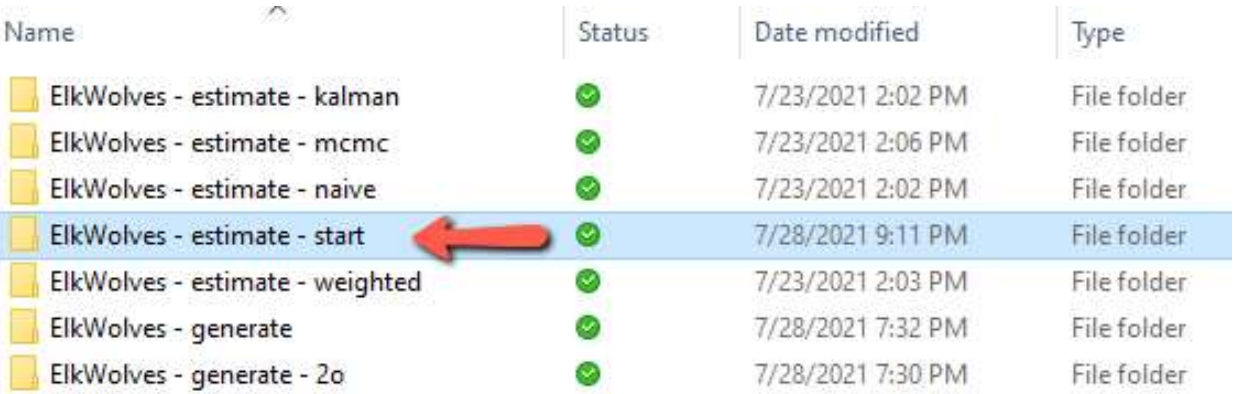
- **A run (.vdfx) with the best parameters**
- **An output file (.out) summarizing the parameters found**
  - Parameters can then be reused by loading the .out as a Changes file (like .cin files)
- **A Payoff Report (.rep) with diagnostics (optionally)**

Naïve Calibration Setup

- **Copy the data files you created from the data generator model to the "start" model folder**

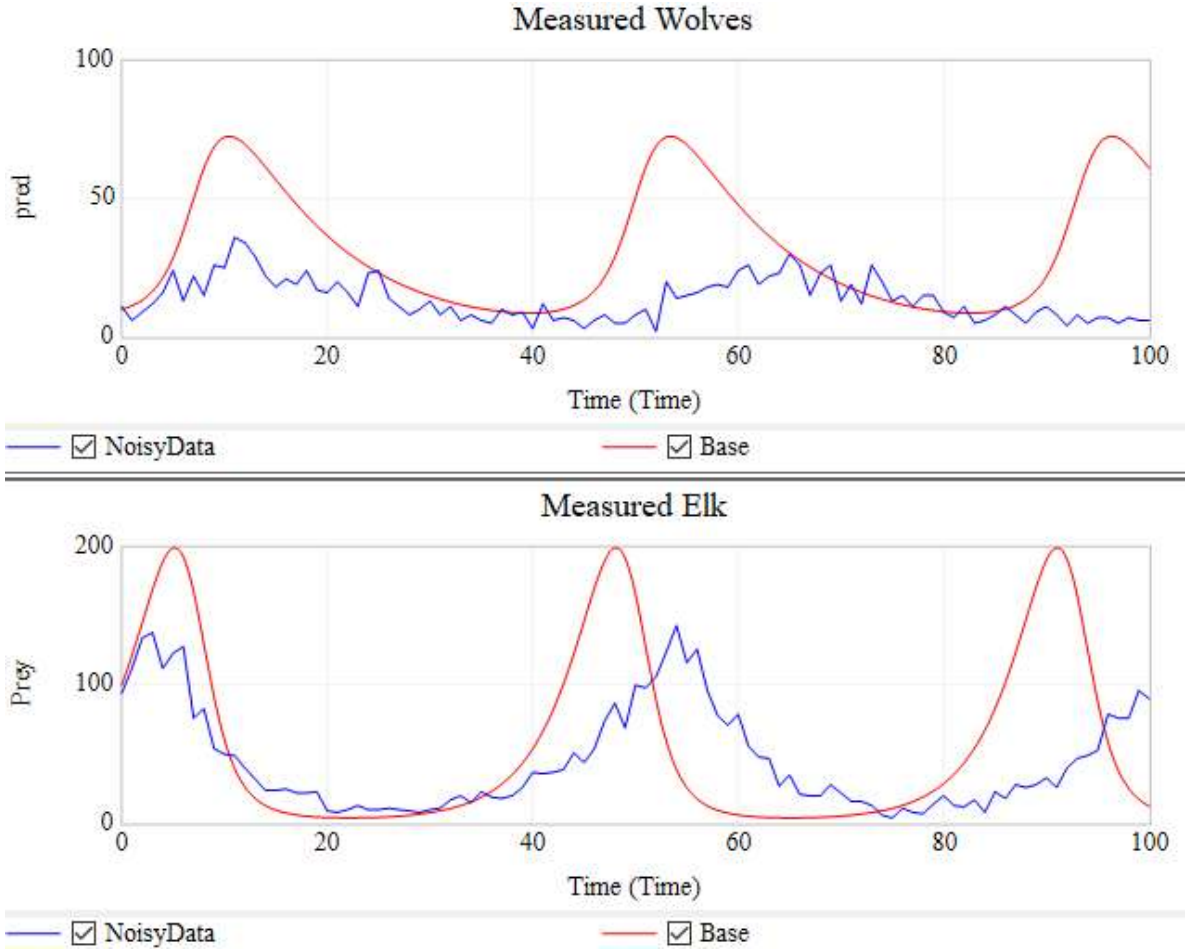


- **Open the starting point model**



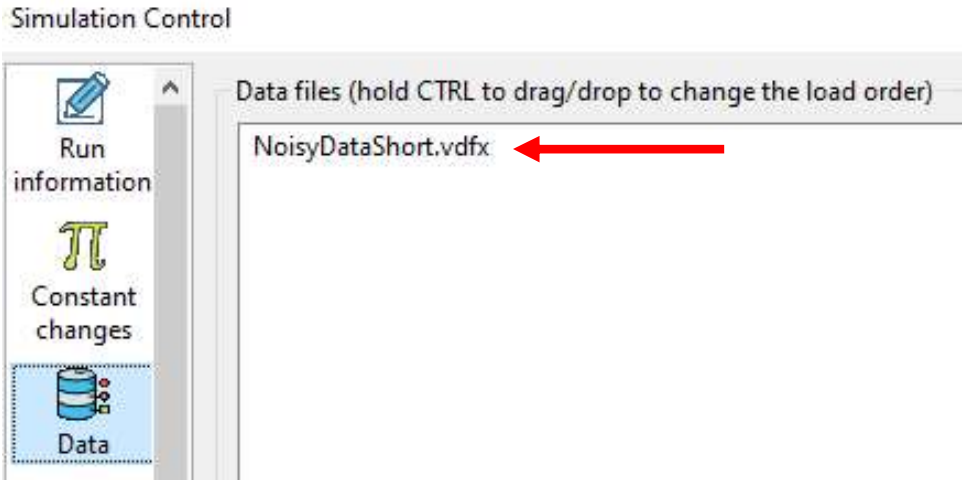
Naïve Calibration Setup

- **Do a "Base" run (uncalibrated)**
- **Load the data from NoisyData or NoisyDataShort**
- **Notice how the model doesn't fit the data well**



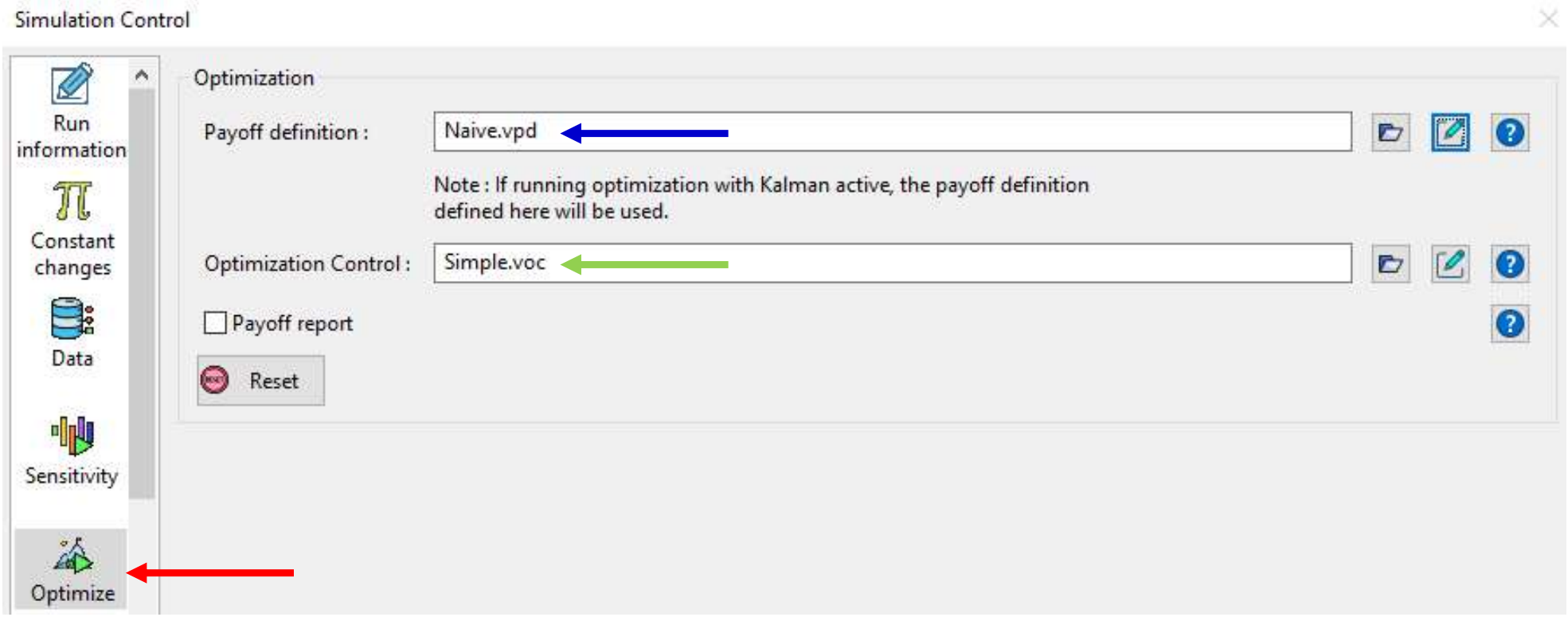
Naïve Calibration Setup

- **Change the runname ("NaiveCal.vdfx" or similar)**
- **Go to the Data tab**
  - Load comparison data (recommend the short version)

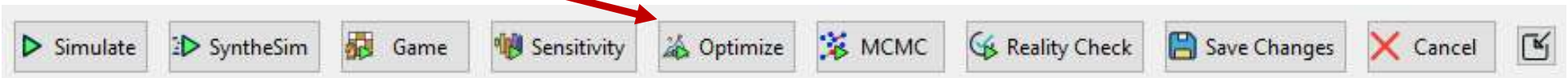


Naïve Calibration Setup (Continued)

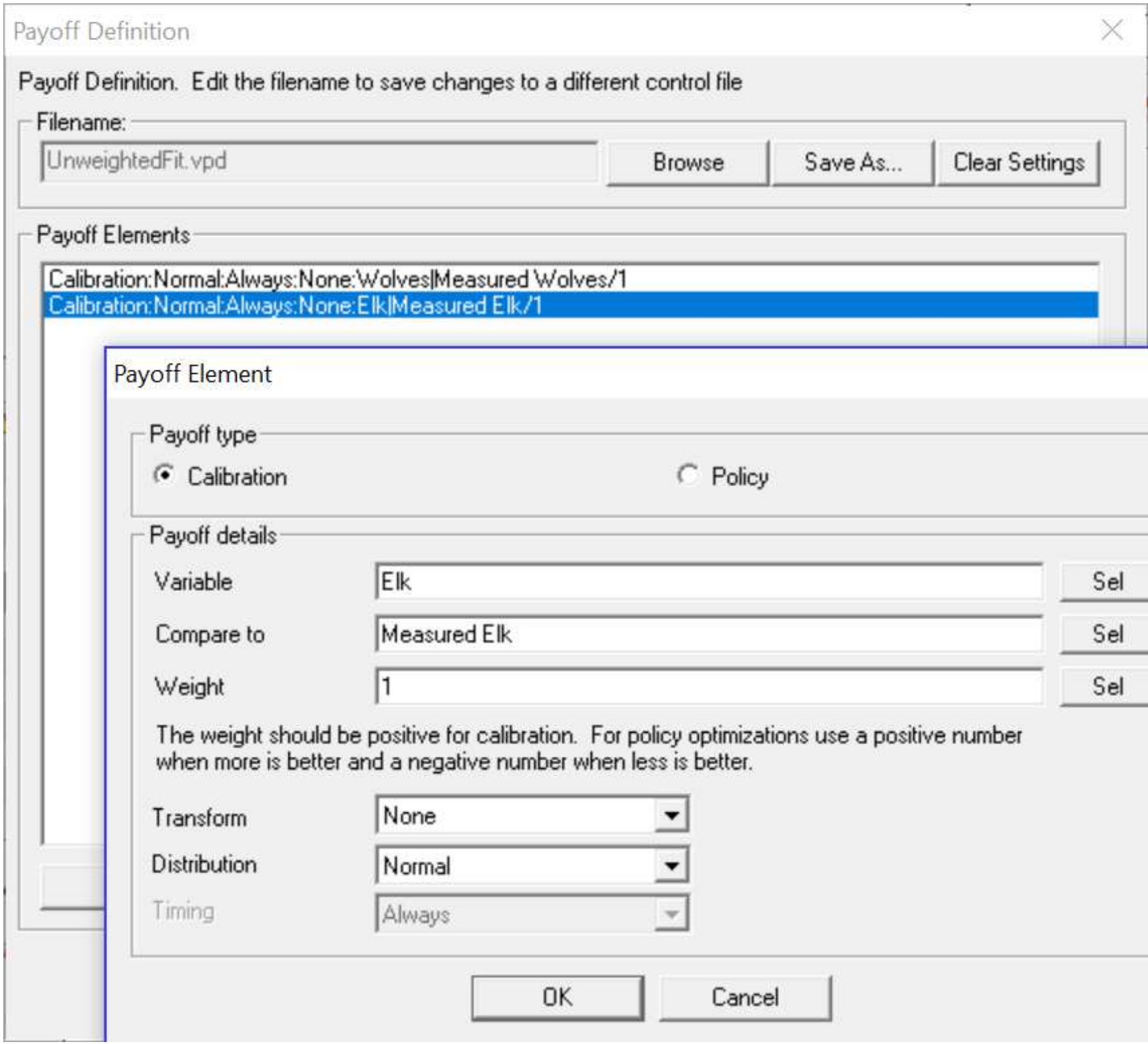
- **Change the runname ("NaiveCal.vdfx" or similar)**
- **Go to the Optimize pane**
  - Create a Payoff (.vpd)
  - Create a Control file (.voc)



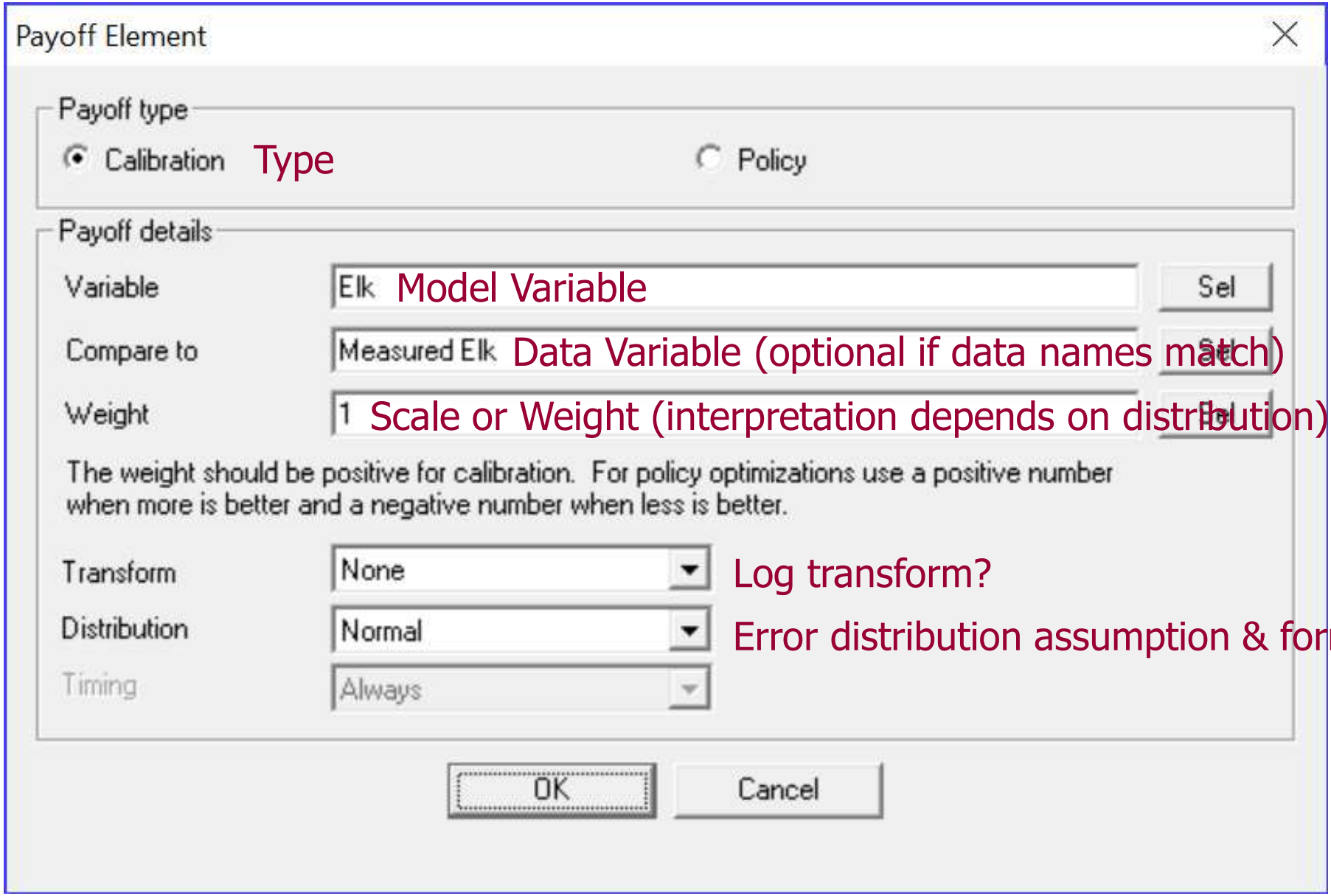
- **Hit the Optimize button**



Payoff (.vpd)


Payoff Definition
Payoff Definition. Edit the filename to save changes to a different control file
Filename:
UnweightedFit.vpd
Browse
Save As...
Clear Settings
Payoff Elements
Calibration:Normal:Always:None:Wolves|Measured Wolves/1
Calibration:Normal:Always:None:Elk|Measured Elk/1
Payoff Element
Payoff type
Calibration
Policy
Payoff details
Variable
Elk
Sel
Compare to
Measured Elk
Sel
Weight
1
Sel
The weight should be positive for calibration. For policy optimizations use a positive number when more is better and a negative number when less is better.
Transform
None
Distribution
Normal
Timing
Always
OK
Cancel

Calibration Payoff Types



The Payoff File (.vpd) as text

```
*C
```

Keyword indicating type (calibration = *C, policy = *P, etc.)

```
wolves|measured wolves/1
```

Model variable | data variable / weight or scale parameter

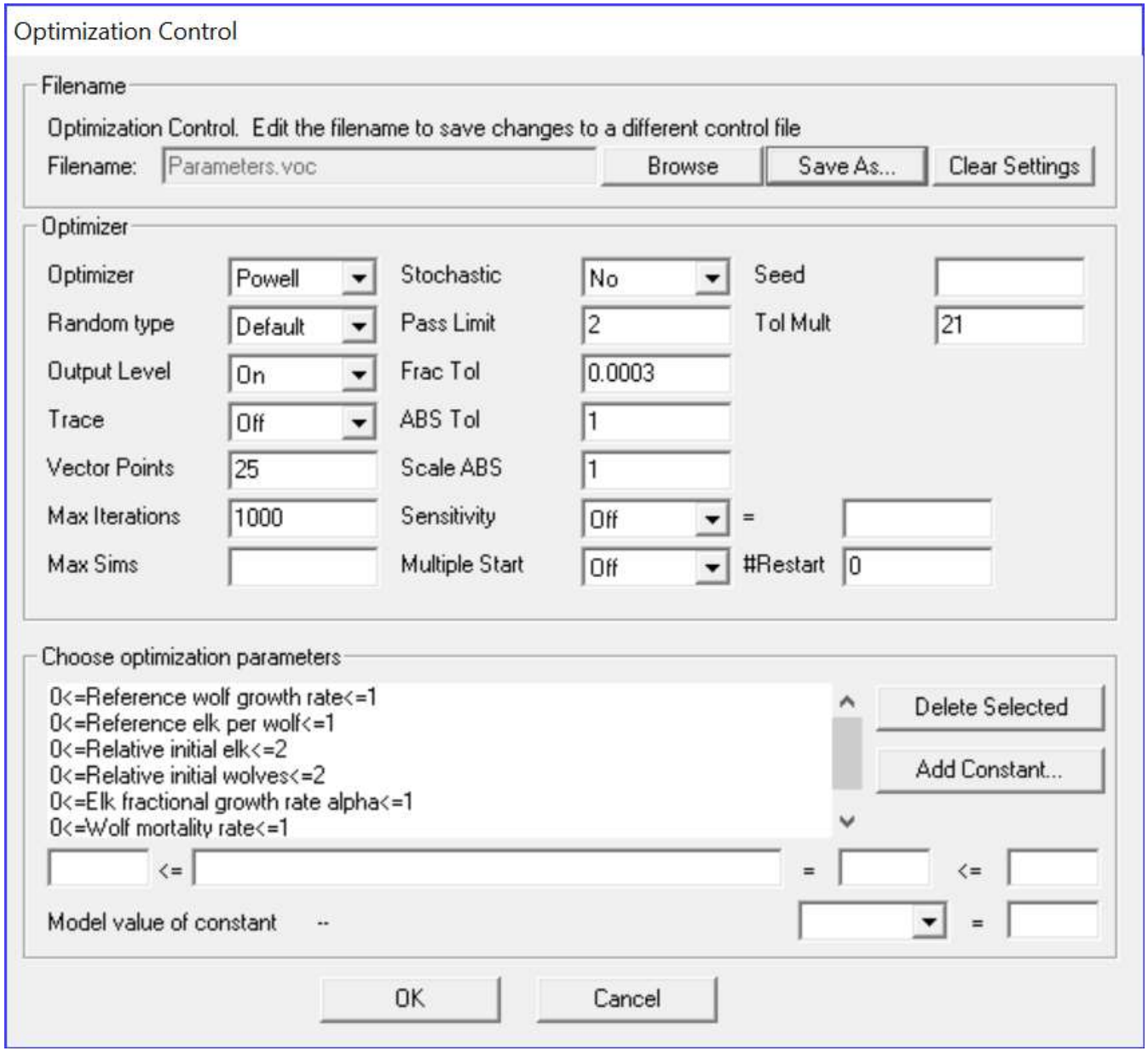
The weight can also be a variable.

Subscript ranges are OK, as long as they match.

Optimization Control File (.voc)

Method & Settings

Parameters & Bounds



The Optimization Control File as text

```
:OPTIMIZER=Powell
:SENSITIVITY=Off
:MULTIPLE_START=Off
```

<bla bla bla – algorithm control settings>

List of parameters to optimize:

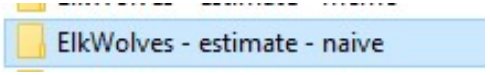
```
0<=Reference wolf growth rate<=1
0<=Reference elk per wolf<=1
0<=Relative initial elk<=2
```

…

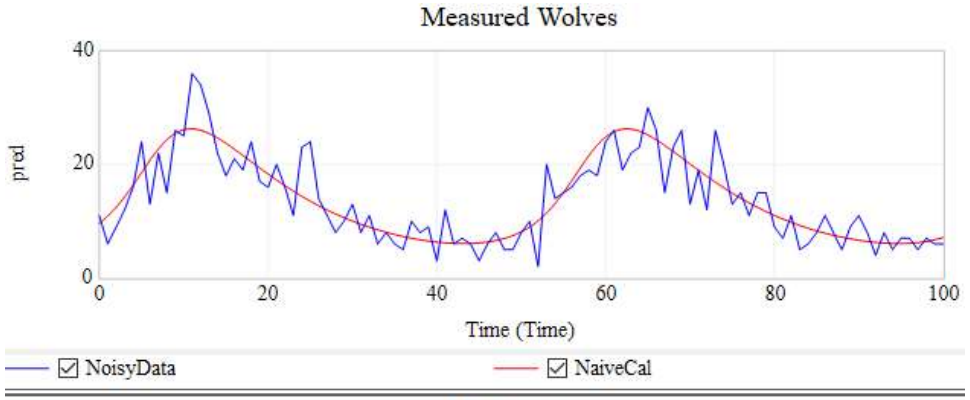
Min <= Variable Name = Initial Guess <= Max

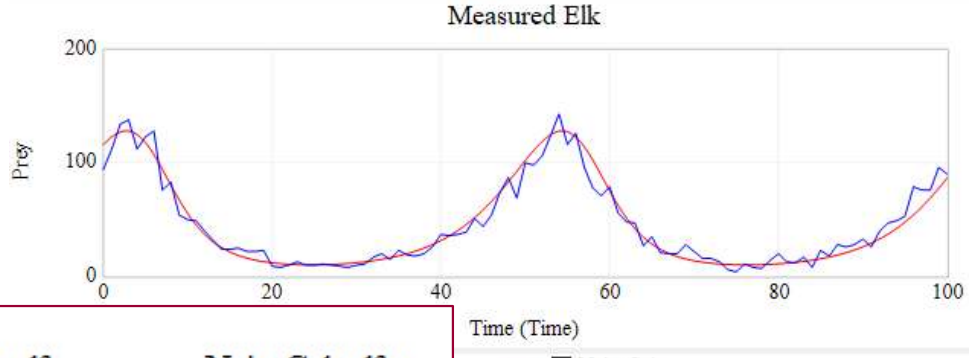
Subscript ranges are OK. Initial Guess is often omitted.

Optimize
Sample results are in



- **The model now fits the data (hopefully)**
- **It fits in the future, after the short data runs out (we probably made this too easy)**





- **Verify: use the Runs Compare tool to see if the parameters match the synthetic data model**

NoisyData.vdfx	NaiveCal.vdfx
0.22	0.22226253
0.2	0
0.2	0
0.16	0.16122551
0.18000001	0.16827042
1.08	1.1602107
0.94999999	0.96516252
0.2	0

Variations

- **Try again with …**
  - an even shorter input data series
  - more noise in the generator
  - the 2nd order model data
  - a longer forecast horizon