

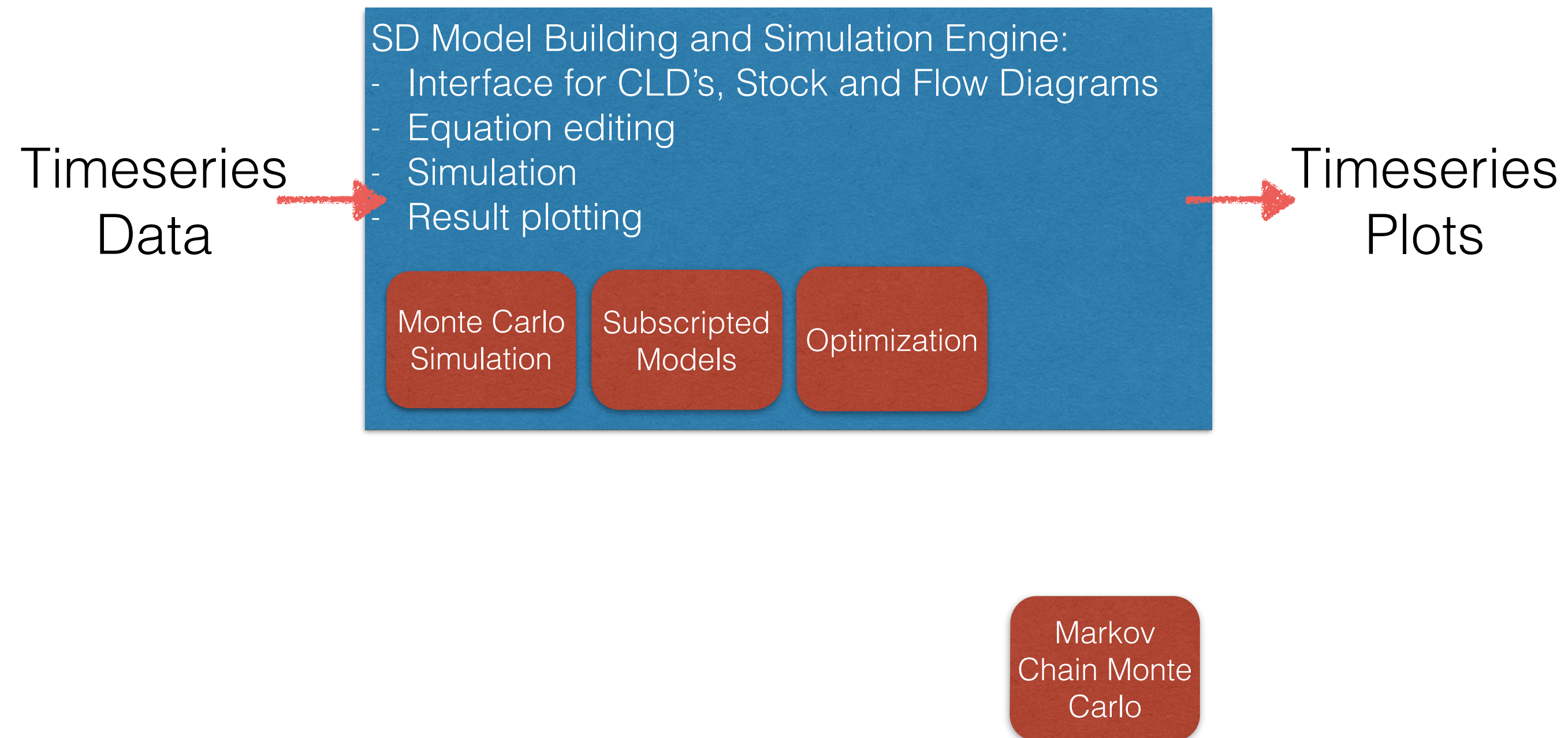
pysd: System Dynamics Models in Python

purpose, capabilities, examples, structure,
development plan

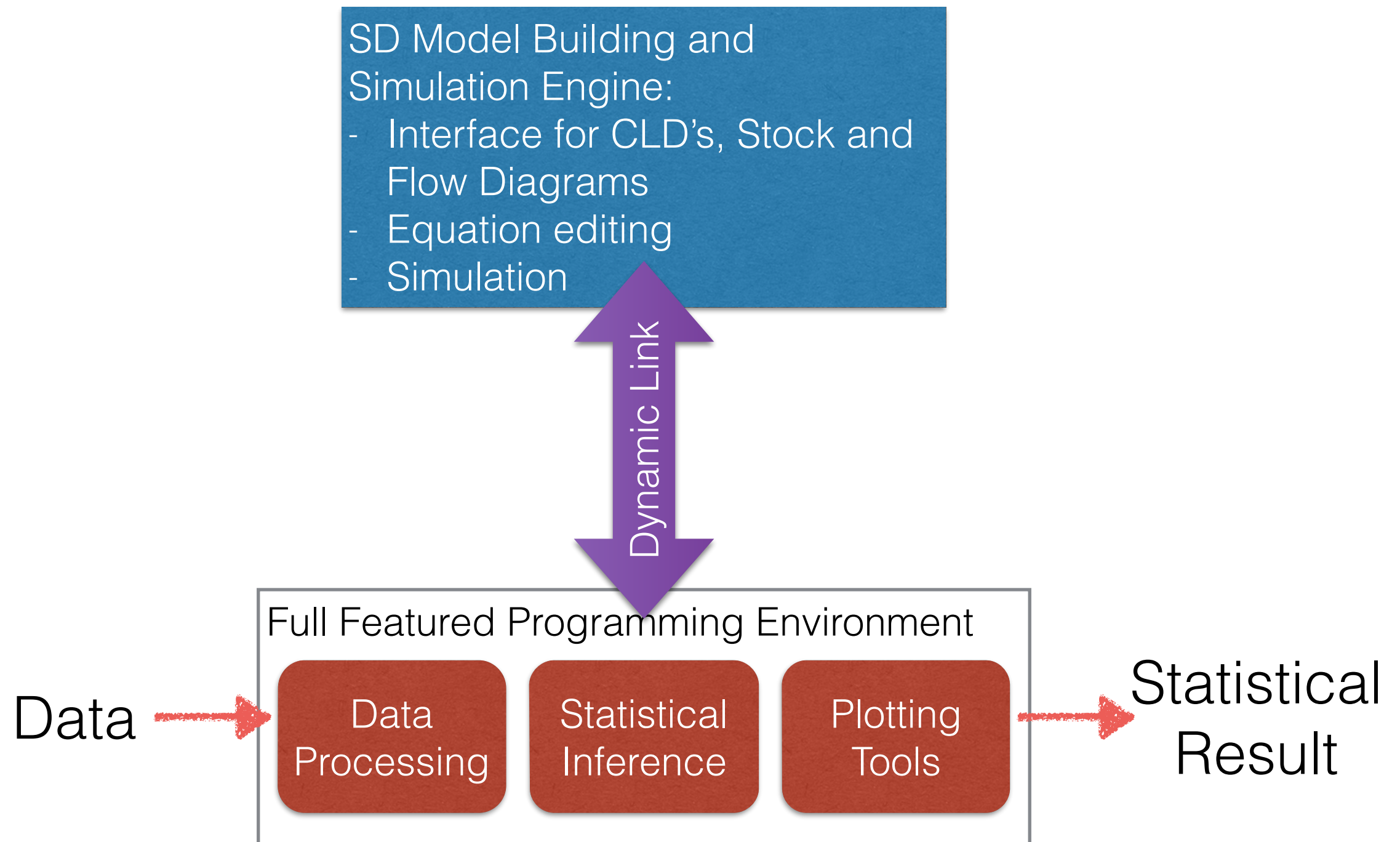
Motivation

- More powerful analysis tools
- Better ways to structure multi-part models
- Tighter integration with “big data” (including non-timeseries data)
- Integration with other simulation and analysis methods
- Better communication of analysis method and results (replicability!)

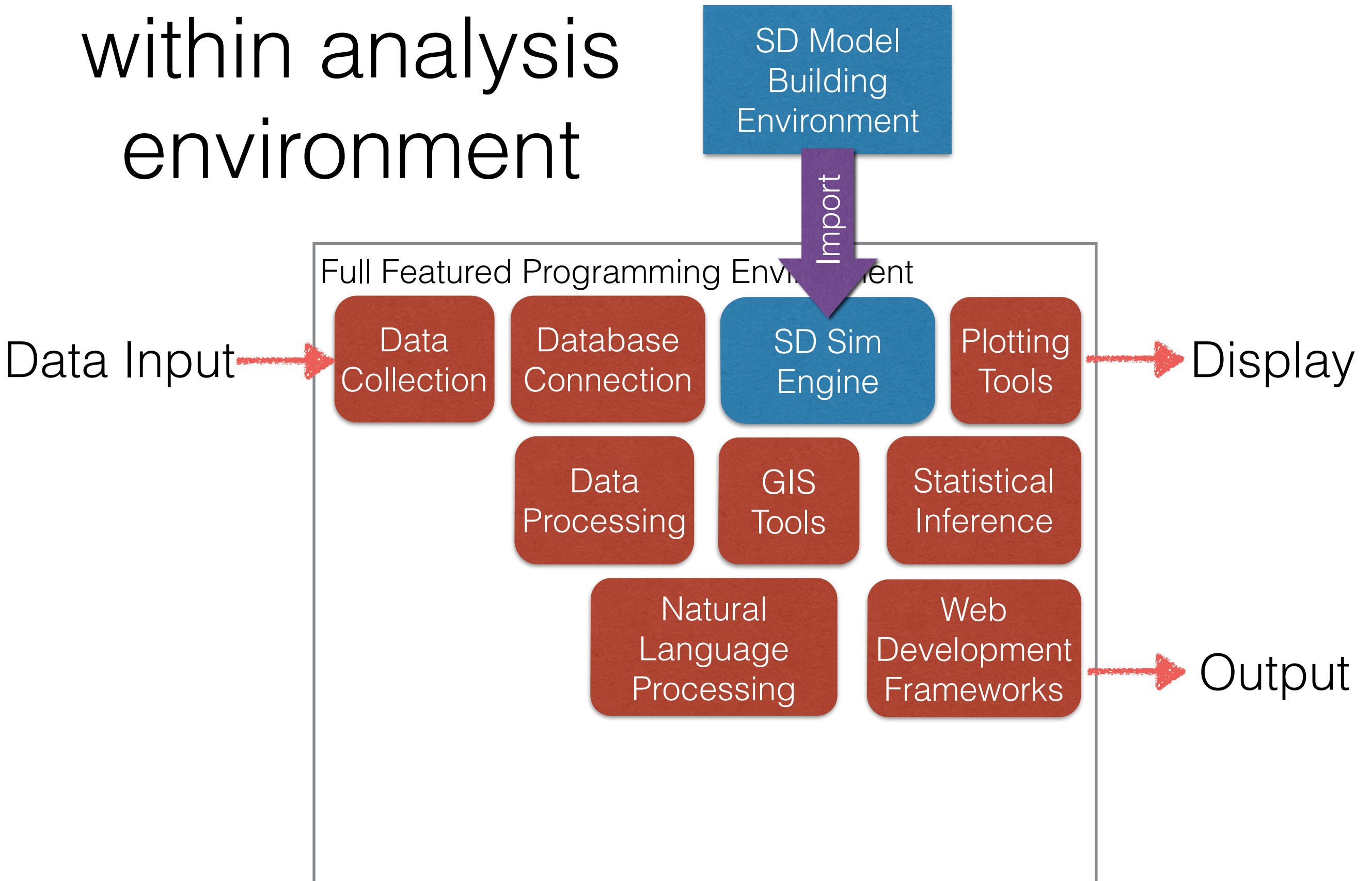
Traditional: Bring analysis capabilities into SD software



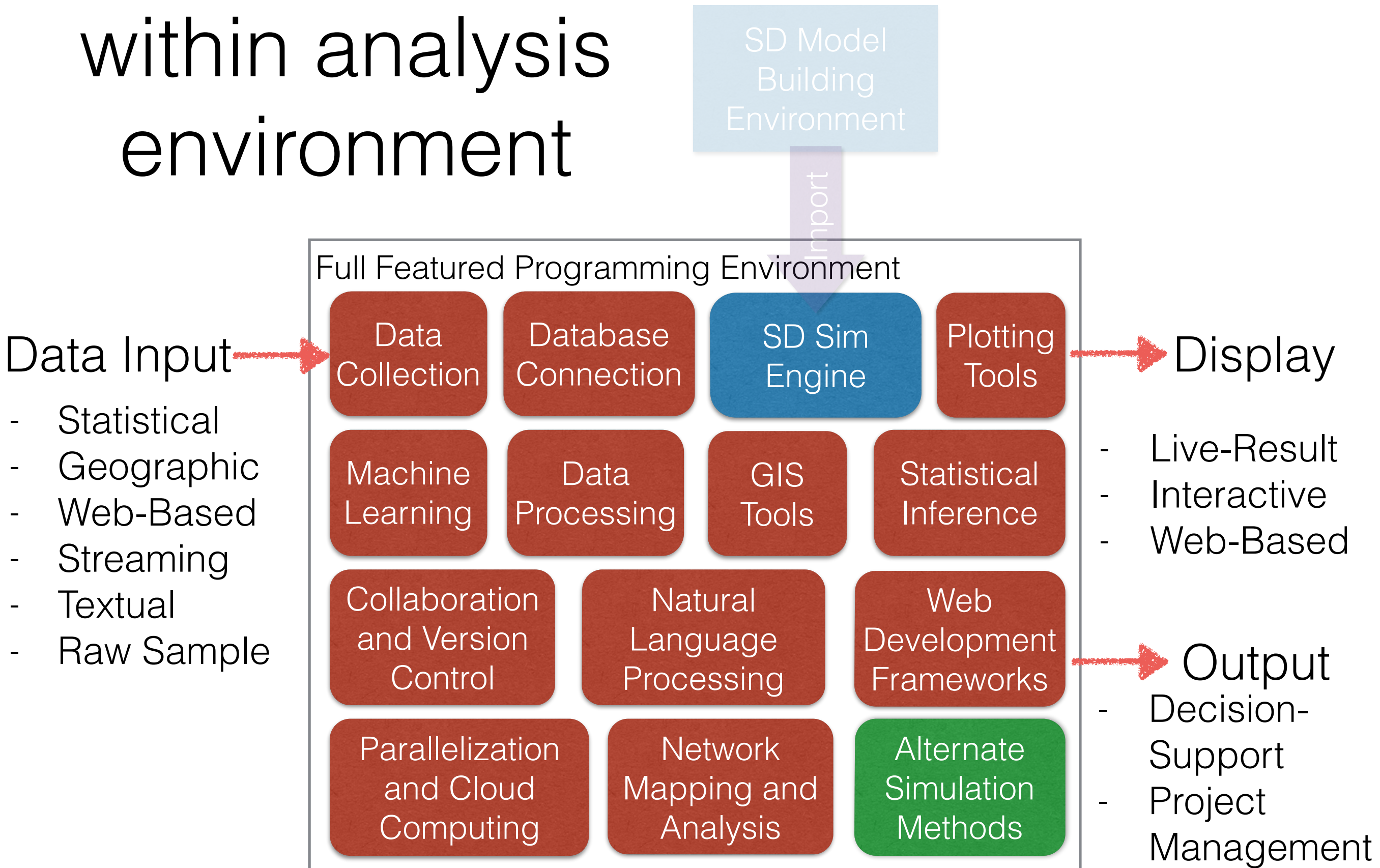
State of the art: Interface SD software with external analysis tools



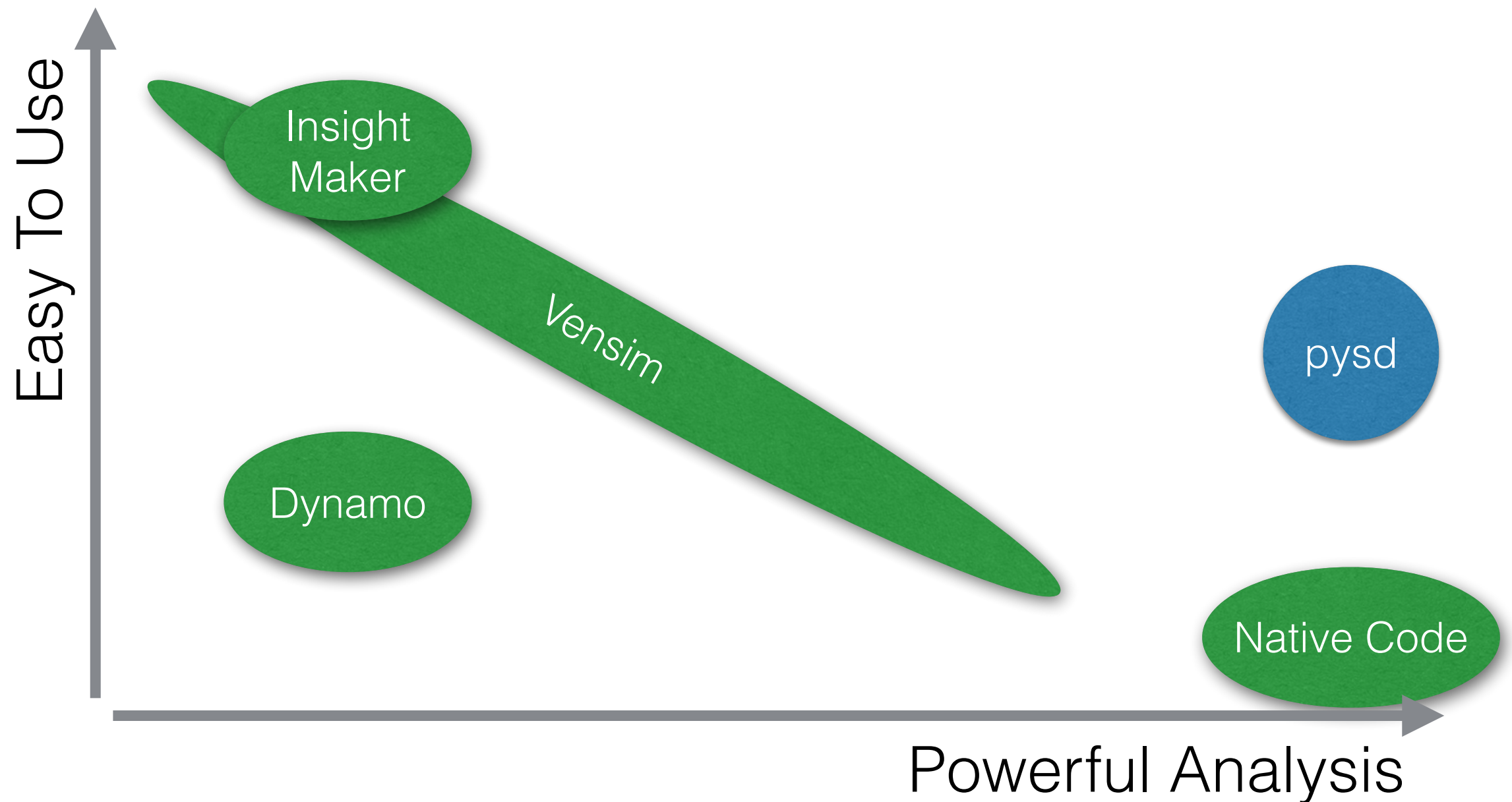
pysd: Simulate within analysis environment



pysd: Simulate within analysis environment



An incomplete map of the SD software space

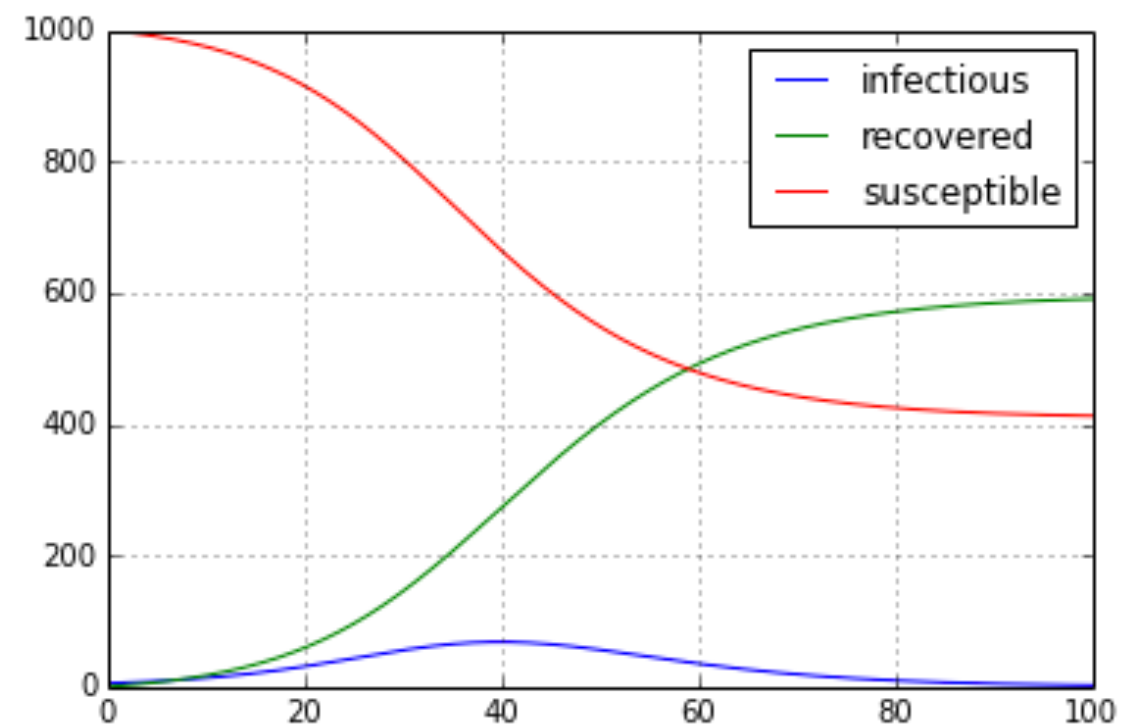
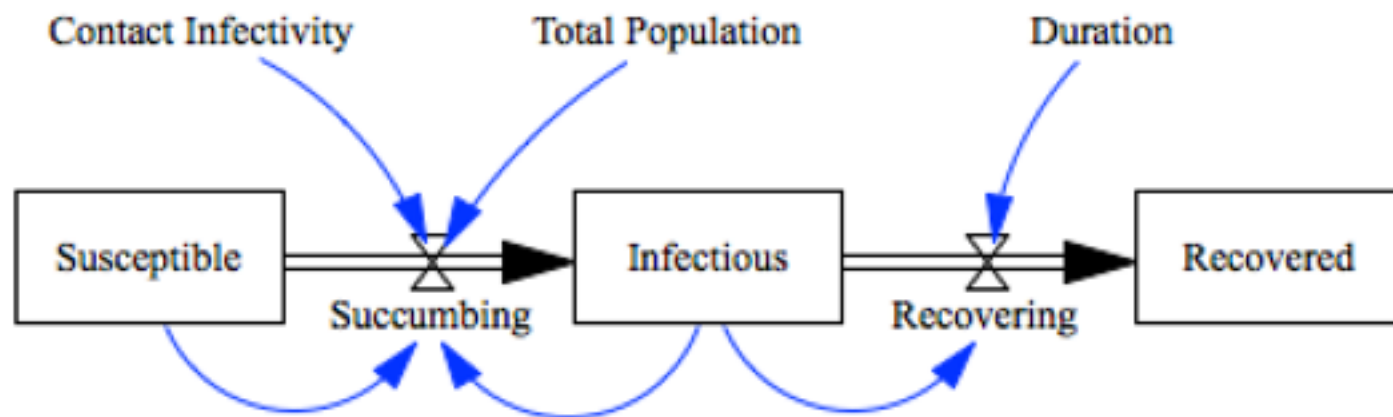


Status

- Functioning prototype available at:
<https://github.com/JamesPHoughton/pysd>
- Supports subset of XMILE, Vensim commands
- Install through python package exchange:
>>> pip install pysd

Basic Usage

```
import pysd
model = pysd.read_vensim('SIR.mdl')
model.run().plot()
```



Mechanics of Importing

SIR.mdl:

Contact Infectivity=
0.3
~ Persons/Persons/Day
~ |

Duration=
5
~ Days
~ |

Infectious= INTEG (
Succumbing-Recovering,
5)
~ Persons
~ |

Recovered= INTEG (
Recovering,
0)
~ Persons
~ |

Recovering=
Infectious/Duration
~ Persons/Day
~ |



```
def model_function(stocks, t):  
    infectious, recovered, susceptible, = stocks
```

```
    total_population = 1000
```

```
    duration = 5
```

```
    recovering = infectious/duration
```

```
    contact_infectivity = 0.3
```

```
    succumbing = susceptible*infectious/  
        total_population*  
        contact_infectivity
```

```
    dinfectious_dt = succumbing-recovering
```

```
    dsusceptible_dt = -succumbing
```

```
    drecovered_dt = recovering
```

```
    return [dinfectious_dt, drecovered_dt,  
            dsusceptible_dt]
```

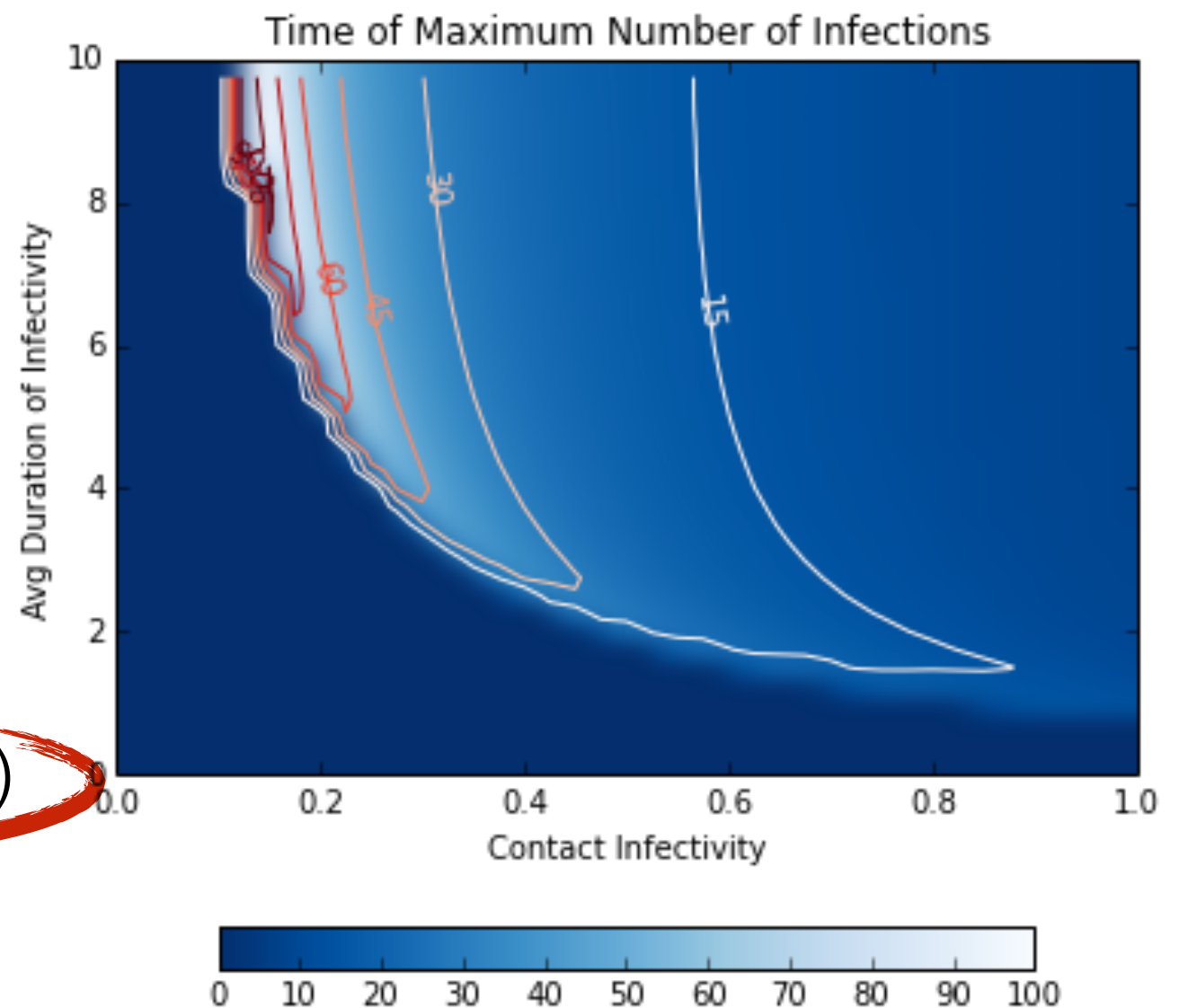
```
odeint(model_function, initial_values, tseries)
```

Model Exploration

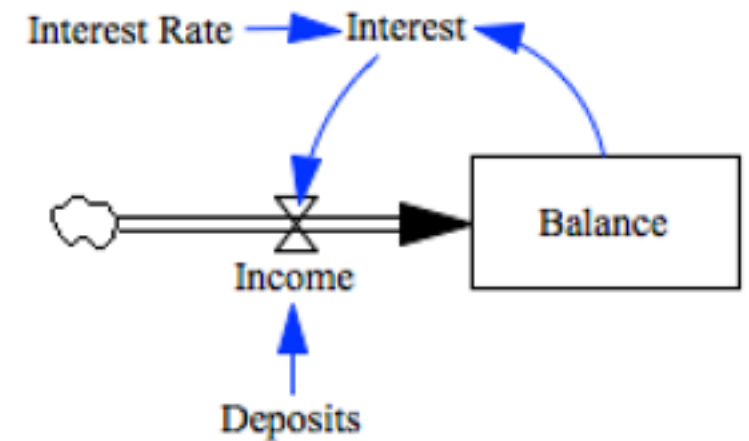
```
ci = np.arange(0,1,.025)  
d = np.arange(.5,10,.25)  
ci_grid, d_grid = np.meshgrid(ci, d)
```

```
def tmax(ci, d):  
    ps = {'contact_infectivity':ci,  
          'duration':d}  
    stocks = model.run(params=ps)  
    return stocks['infectious'].idxmax()
```

```
vtmax = np.vectorize(tmax)  
tmax_grid = vtmax(ci_grid, d_grid)
```



Parallelization



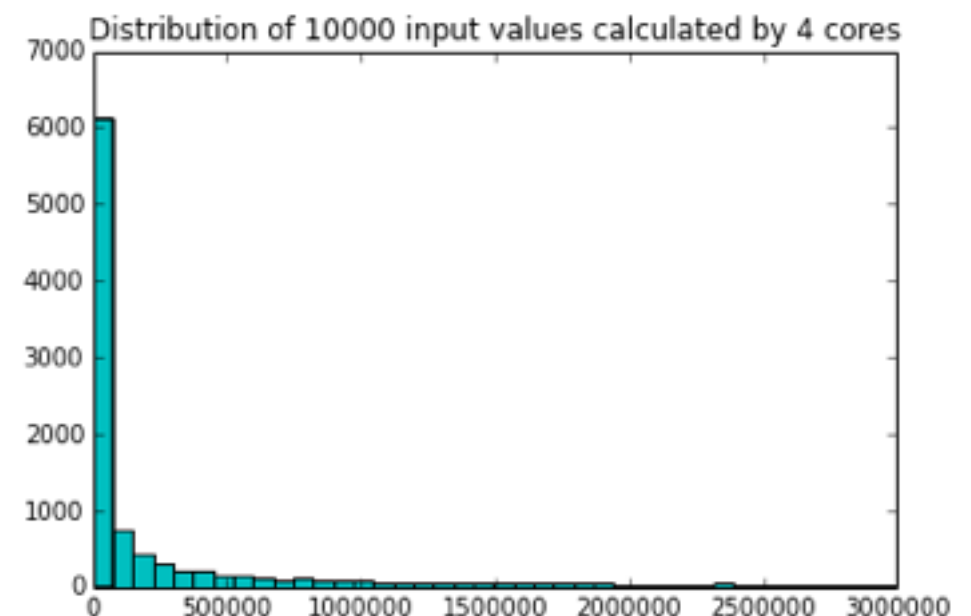
```
model = pysd.read_vensim('bank_balance.mdl')
interest_rate = np.random.uniform(high=.1, size=10000)
```

```
def final(rate):
    ps={'interest_rate':rate}
    stocks = model.run(params=ps)
    return stocks['balance'].iloc[-1]
```

Series Monte Carlo:

```
[final(rate) for rate in interest_rate]
```

Execution Time: 11s

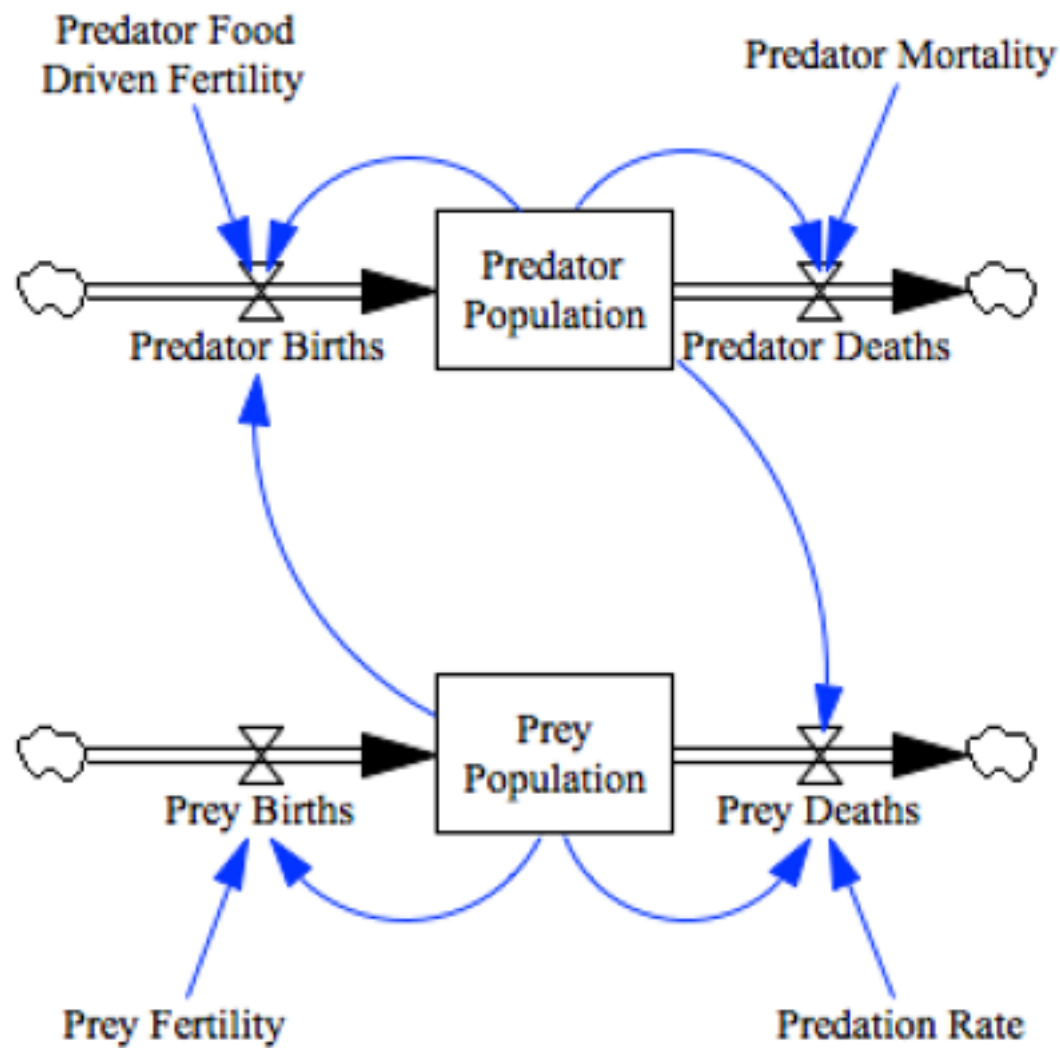


Parallel Monte Carlo:

```
from IPython.parallel import Client
dview = Client().cli[:]
dview.push(model)
dview.map(final, interest_rate)
```

Execution Time: 2.9s

Non-Timeseries Optimization



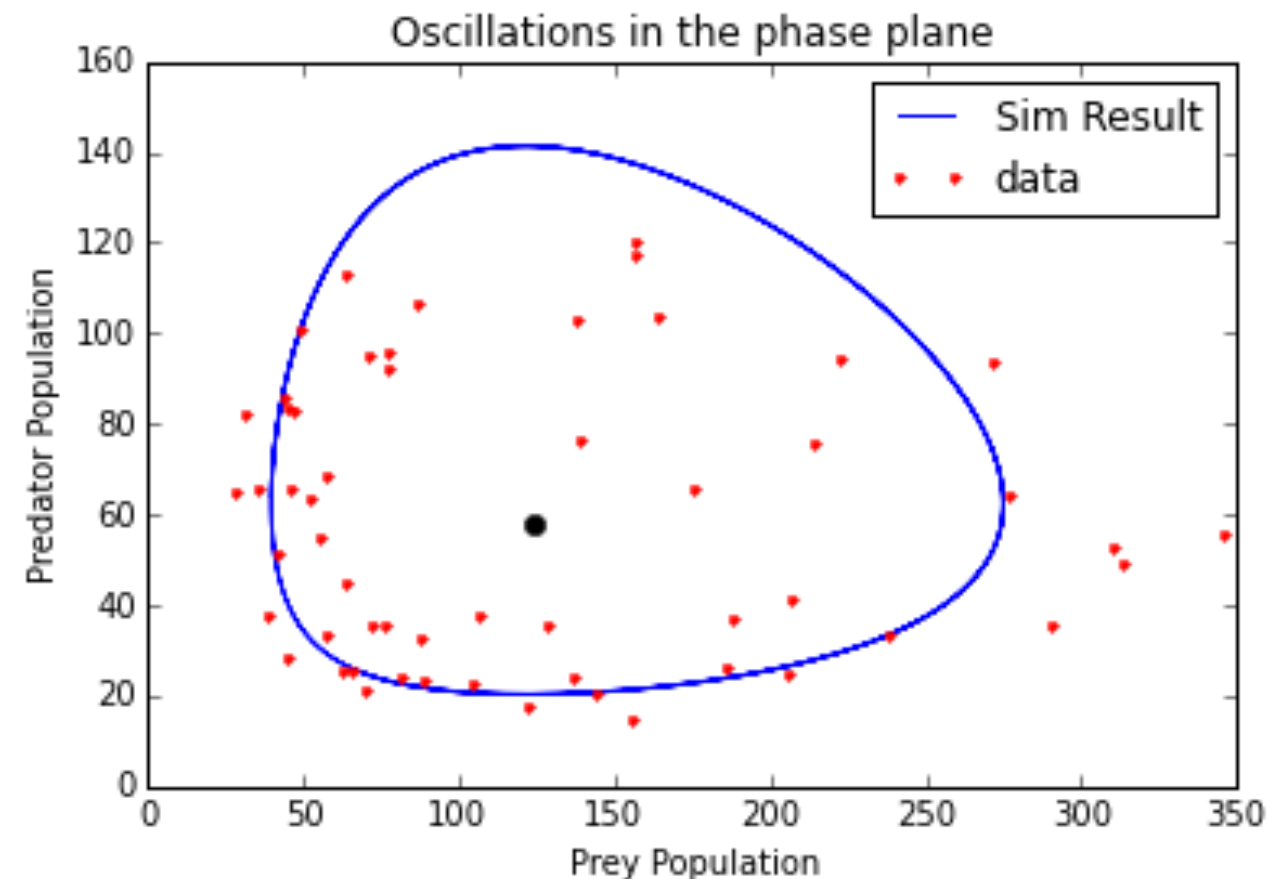
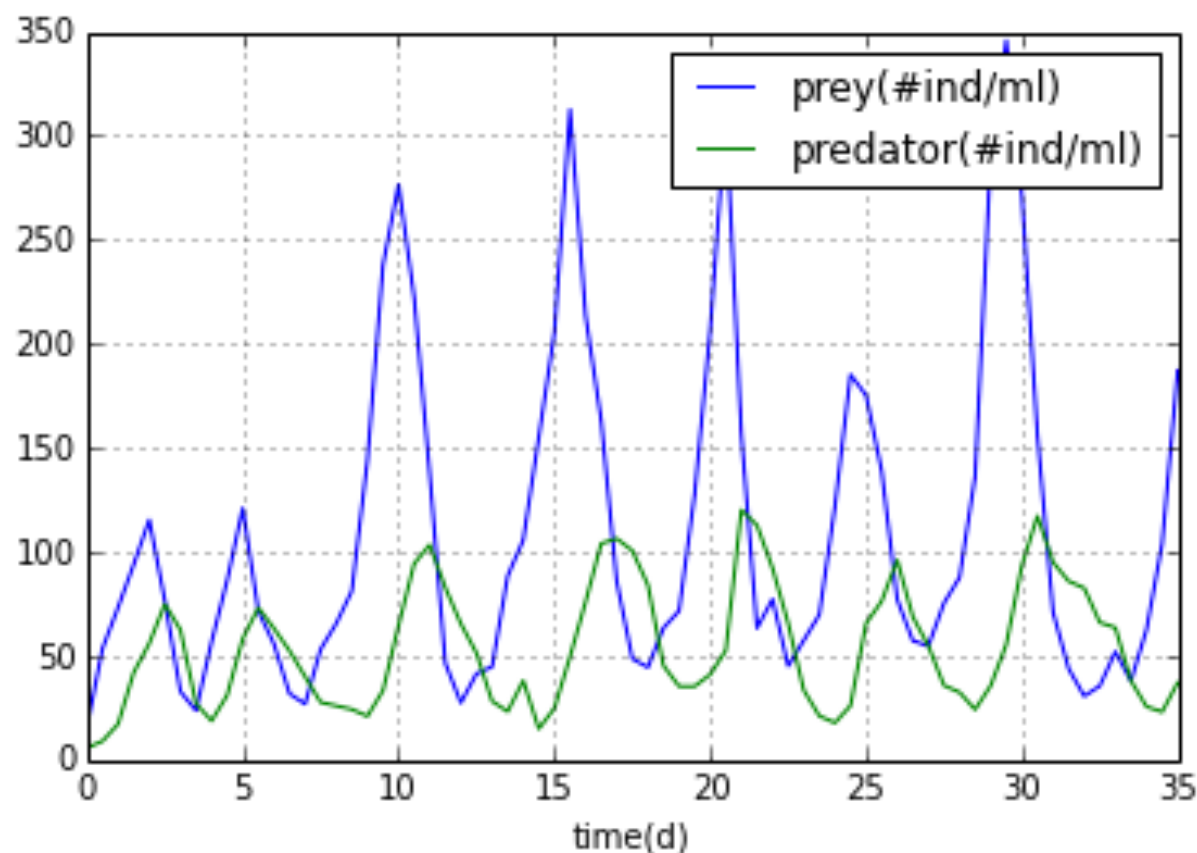
```
def error(parameter_list):
```

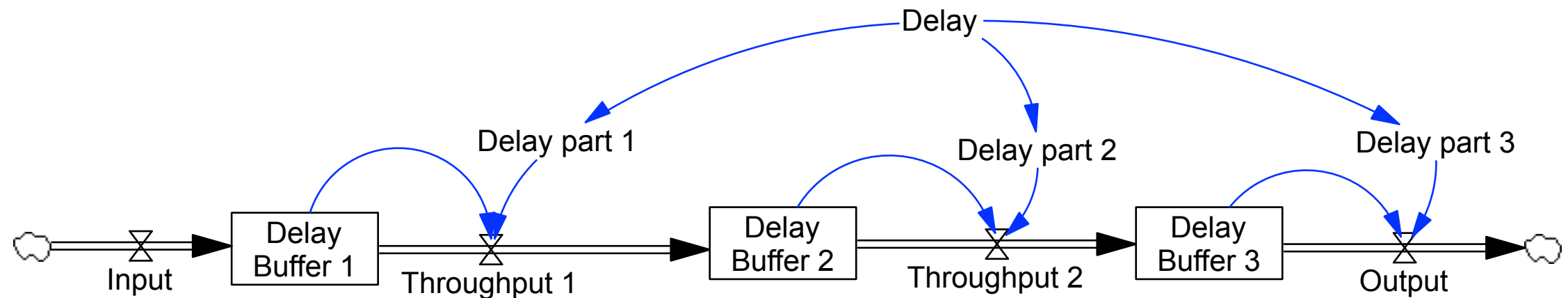
```
...
```

```
errors = sim_result - data_transform
phase_plot_error = (errors**2).sum()
```

```
...
```

```
scipy.optimize.minimize(error, x0=[.005, 1, 1, .002],
                        method='L-BFGS-B',
                        bounds=[(0,10), (0,None),
                                (0,10), (0,None)])
```





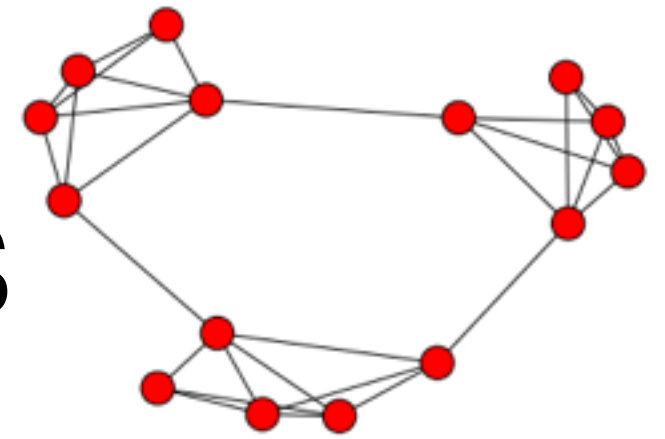
```
def animate(t):
    #draw a live result
    ...
    #run the simulation forward
    global stocks
    model.tstart = x[-1]
    model.tstop = x[-1]+1./fps
    model.initial_values = stocks
    ps={'input':input_val,
        'delay':adjustment_delay})
    stocks = model.run(params=ps).iloc[-1]
    ...
    #collect user input
    ...
```

```
anim = animation.FuncAnimation(fig, animate,
                                frames=seconds*fps)
```

Realtime Interaction

Maybe a demo?

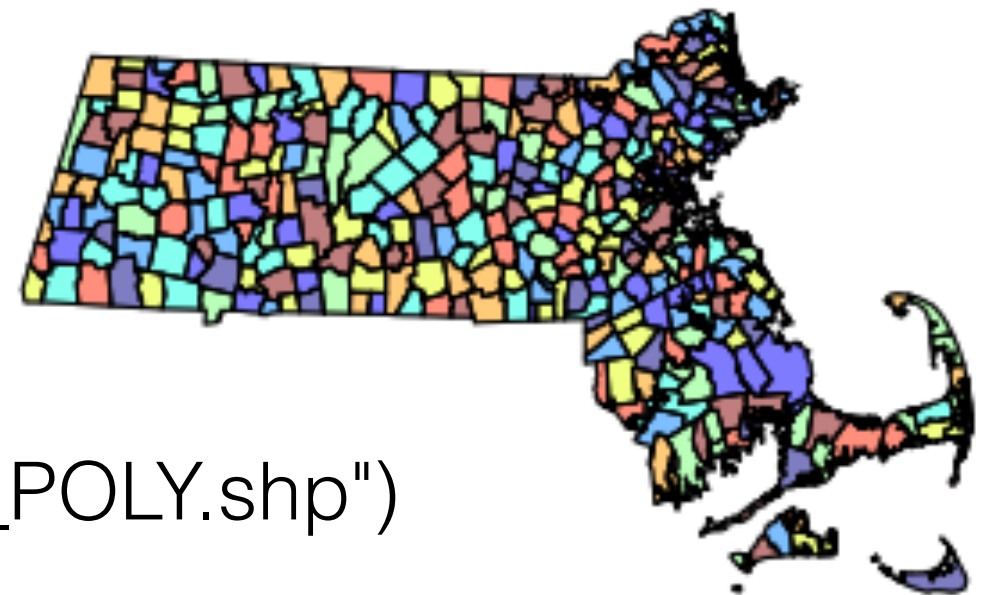
Network/Patch Models



```
import networkx as nx  
g = nx.graph(...)
```

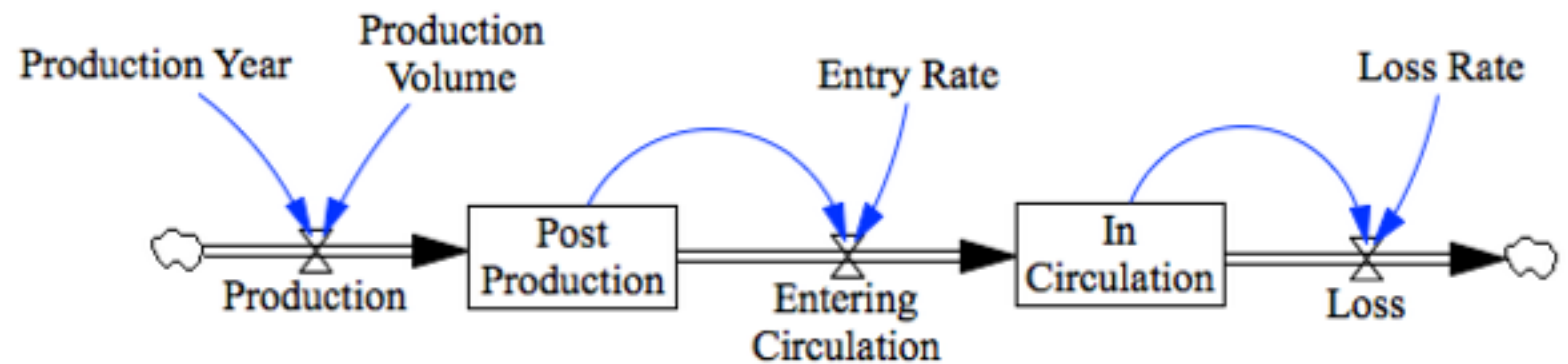
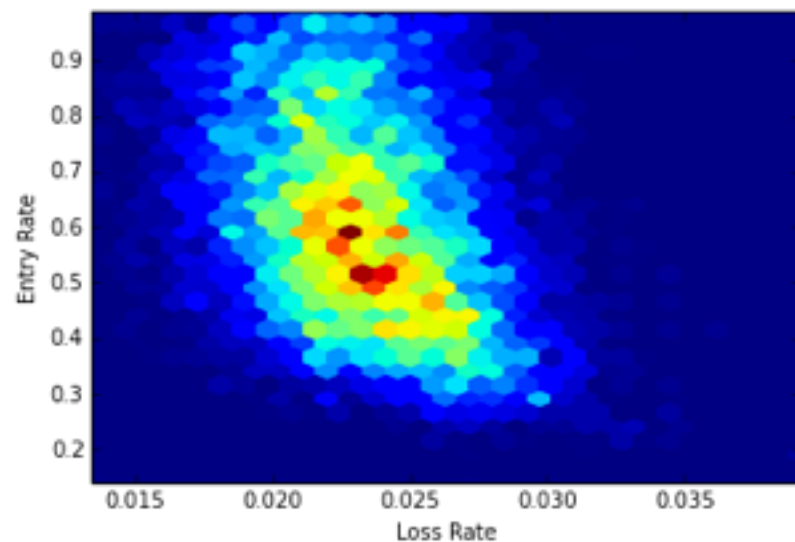
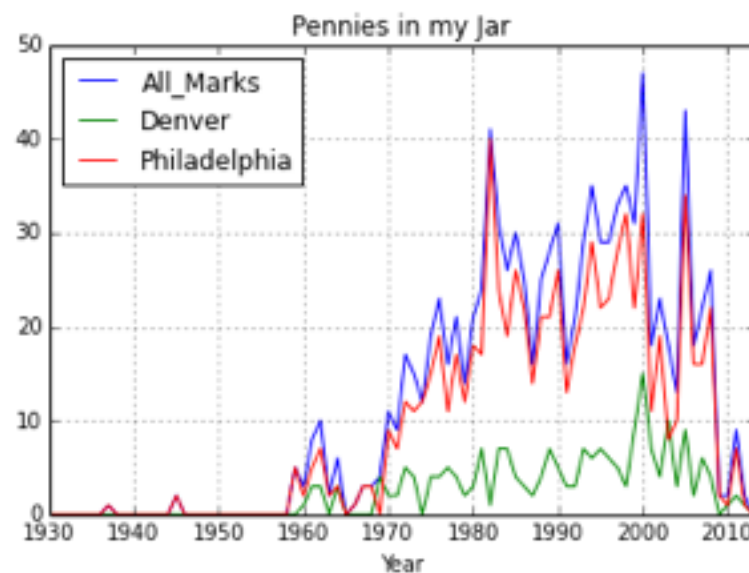
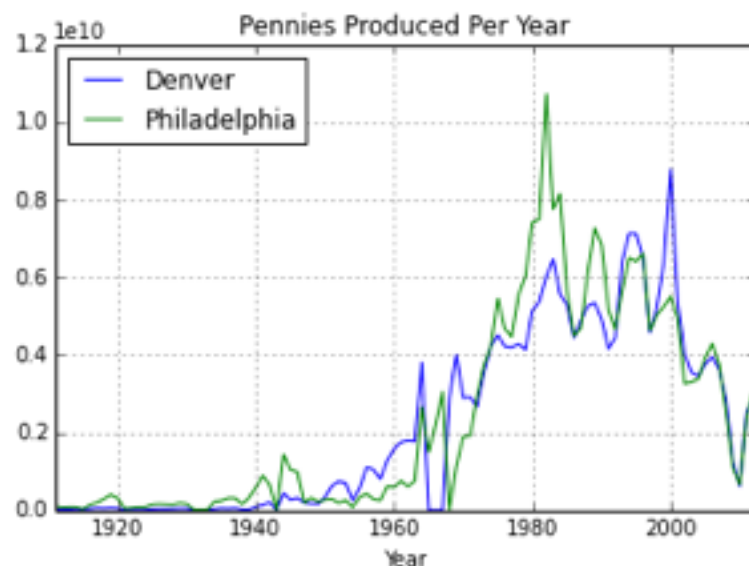
```
for node in g.nodes_iter():  
    g[node]['model'] = pysd.read_vensim('Patch SIR.mdl')
```

```
for t in np.arange(model.tstart, model.tstop, model.dt):  
    for node in g.nodes_iter():  
        ps = {'cross_infectivity': nx.edgelist(node)... }  
        g[node]['model'].run(params=ps)
```



```
import geopandas as gpd  
towns = gpd.read_file("TOWNSSURVEY_POLY.shp")
```

Subscripted MCMC



```
models = [[year, pysd.read_vensim('penny_jar.mdl')]
           for year in range(1930,2014)]
```

```
entry_rate = mc.Uniform('entry_rate', lower=0,
                        upper=.99, value=.08)
```

```
loss_rate = mc.Uniform('loss_rate', lower=0,
                       upper=.3, value=.025)
```

```
@mc.stochastic(trace=True, observed=True)
```

```
def circulation(...)
```

```
    population = models['model'].run(...)
```

```
    ...
```

```
    return log_prob
```

```
mcmc = mc.MCMC(...)
```

```
mcmc.sample(20000)
```

To be demonstrated:

- Fitting models to streaming data
- Machine learning regressions in place of lookups
- Reversible jump MCMC for model selection
- Web-based interactive result display
- Runtime connection to ABM, etc...
- Adaptive sampling of parameter spaces
- Integration with decision models
- ...

To be Developed:

- cython/theano speedups
- Step function with memory
- Smart parameter modification
- Additional XMILE/Vensim translation ability