# minimal\_prey\_predator

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## 1 Prey-Predator simulation in cadCAD

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This is an cadCAD simulation of the Lotka-Volterra prey-predator model. This is a minimal example contained inside a notebook, and it can be used for quick experimentations

## 1.1 Dependences

```
[1]: %%capture !pip install cadcad
```

```
[2]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
from cadCAD.configuration import Experiment
from cadCAD.configuration.utils import config_sim
from cadCAD.engine import ExecutionMode, ExecutionContext, Executor
```

### 1.2 Definitions

## 1.2.1 Initial conditions and parameters

```
[3]: initial_conditions = {
        'prey_population': 100,
        'predator_population': 15
     }

params = {
        "prey_birth_rate": [1.0],
        "predator_birth_rate": [0.01],
        "predator_death_const": [1.0],
        "prey_death_const": [0.03],
        "dt": [0.1] # Precision of the simulation. Lower is more accurate / slower
}

simulation_parameters = {
```

```
'N': 7,
'T': range(200),
'M': params
}
```

#### 1.2.2 Policies

```
[4]: def p_predator_births(params, step, sL, s):
       dt = params['dt']
       predator_population = s['predator_population']
      prey_population = s['prey_population']
       birth_fraction = params['predator_birth_rate'] + np.random.random() * 0.0002
       births = birth_fraction * prey_population * predator_population * dt
       return {'add_to_predator_population': births}
     def p_prey_births(params, step, sL, s):
       dt = params['dt']
      population = s['prey_population']
      birth_fraction = params['prey_birth_rate'] + np.random.random() * 0.1
      births = birth_fraction * population * dt
       return {'add_to_prey_population': births}
     def p_predator_deaths(params, step, sL, s):
       dt = params['dt']
      population = s['predator_population']
       death_rate = params['predator_death_const'] + np.random.random() * 0.005
       deaths = death_rate * population * dt
       return {'add_to_predator_population': -1.0 * deaths}
     def p_prey_deaths(params, step, sL, s):
      dt = params['dt']
       death_rate = params['prey_death_const'] + np.random.random() * 0.1
      prey_population = s['prey_population']
      predator_population = s['predator_population']
       deaths = death_rate * prey_population * predator_population * dt
       return {'add_to_prey_population': -1.0 * deaths}
```

## 1.2.3 State update functions

```
[5]: def s_prey_population(params, step, sL, s, _input):
    y = 'prey_population'
    x = s['prey_population'] + _input['add_to_prey_population']
    return (y, x)
```

```
def s_predator_population(params, step, sL, s, _input):
    y = 'predator_population'
    x = s['predator_population'] + _input['add_to_predator_population']
    return (y, x)
```

## 1.2.4 State update blocks

## 1.2.5 Configuration and Execution

```
Execution Mode: local_proc
Configuration Count: 1
Dimensions of the first simulation: (Timesteps, Params, Runs, Vars) = (200, 5, 7, 2)
Execution Method: local_simulations
SimIDs : [0, 0, 0, 0, 0, 0, 0]
SubsetIDs: [0, 0, 0, 0, 0, 0, 0]
Ns : [0, 1, 2, 3, 4, 5, 6]
ExpIDs : [0, 0, 0, 0, 0, 0, 0]
Execution Mode: parallelized
Total execution time: 0.15s
```

## 1.2.6 Results

```
[8]: import plotly.express as px
```