# SMC Simulating models deterministically

#### September 2, 2022

This notebook deterministically simulates the Exclusive Infection, Replacement Infection, Super-infection, Full-Coinfection Uni and Bi-directional conversion models over the same parameters as Spickanll (2013). The models are run for 150 years starting with a population of 998 susceptible people, 1 person infected with the ABS strain and 1 person infected with the ABR strain. For all the model, but for the superinfection and single-strain model, the equilibrium prevalence as derived through stable point analyses has also being plotted.

Note These simulation reveals that the x-axis range in many of Spicknall et al (2013) sub-figures relating model endemic equilibrium prevalence is misprinted (A-C of figure 1 from Spickanll (2013)). The range in the proportion of the population receiving an antibiotic ( $\epsilon$ ) must be higher than 0.15 for these sub-figures.

```
[1]: # Import necesary packages:
   import matplotlib.pyplot as plt
   import os
   import string
   import pandas as pd
   import seaborn as sns
   import numpy as np
   from tqdm import tqdm
   # Import custom module
   import SMB_Code_models_equilibria as models_equil
```

```
[3]: # Empty List for dataframes to concatanate into a single dataframe deterministic_sims = []
# Setup class colors for plotting later.
```

```
[4]: # Function to run a model through
     def run_model_for_epsilon_values(model, param_values, epsilon_vals_as_percent,_
      →time range):
         11 11 11
         Run model with parameter values through a range of % people receiving \Box
      \hookrightarrow treatment.
         Reutrns dataframe of results for the last day of time range.
         data_dict = {'% Receiving Treatment':[],
                      'Class':[],
                      'Individuals':[]
         state_index dict = {state.ID: index for index, state in enumerate(model.
      →state_list)}
         init_state = np.zeros(model.num_state)
         init_state[state_index_dict['S']] = int(param_values ['N']-2)
         init_state[state_index_dict['I_W']] = int(1)
         init_state[state_index_dict['I_Z']] = int(1)
         for percent in tqdm(epsilon_vals_as_percent, desc='% Receiving Treatment'):
             param_values['epsilon']=percent/100
             model.parameters = param_values
             model. initial_values = ( init_state , time_range[0])
             det_run = model. integrate (time_range[1:])
             all_infecteds = {'All Sensitive Infections': [],
                              'All Resistant Infections': []}
             for state, index in state_index_dict.items():
                 value = det_run[-1, index]
                 data_dict['Individuals'].append(value)
                 data_dict['Class'].append(state)
                 data_dict['% Receiving Treatment'].append(percent)
                 if state in ['I_W', 'I_WZ', 'I_WW']:
                     all_infecteds['All Sensitive Infections'].append(value)
                 if state in ['I_Z', 'I_WZ', 'I_ZZ']:
                     all_infecteds['All Resistant Infections'].append(value)
             for state, to_sum in all_infecteds.items():
                 value = np.sum(to_sum)
                 data_dict['Individuals'].append(value)
                 data_dict['Class'].append(state)
                 data_dict['% Receiving Treatment'].append(percent)
         data_df = pd.DataFrame(data_dict)
         full_state_descriptions = {'S': 'Susceptible',
```

```
'I_W': 'Singly Infected Sensitive',

'I_Z': 'Singly Infected Resistant',

'I_WZ': 'Superinfected',

'I_WW': 'Double Infected Sensitive',

'I_ZZ': 'Double Infected Resistant'}

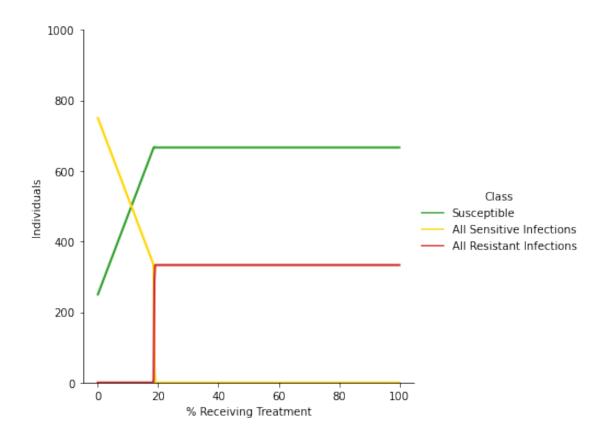
data_df.Class.replace(full_state_descriptions, inplace=True)

return data_df
```

```
[5]: # function for calculating equilria
     def run_equilibria_func_for_epsilon_values(equilibria_func, param_values,_
      →epsilon vals as percent):
         Determine stable equilibria for parameter values through a range of \% people.
      \rightarrow receiving treatment.
         Reutrns dataframe of results.
         data_dict = {'% Receiving Treatment':[],
                      'Class':[],
                      'Individuals':[]
                      }
         for percent in tqdm(epsilon_vals_as percent, desc='% Receiving Treatment'):
             param_values['epsilon']=percent/100
             prevs = equilibria_func(param_values)
             for state, value in prevs[0].items():
                 data_dict['Individuals'].append(value)
                 data_dict['Class'].append(state)
                 data_dict['% Receiving Treatment'].append(percent)
         data df = pd.DataFrame(data dict)
         full_state_descriptions = {'S': 'Susceptible',
                                     'I W': 'All Sensitive Infections',
                                     'I_Z': 'All Resistant Infections'}
         data_df.Class.replace(full_state_descriptions, inplace=True)
         data_df.Individuals = data_df.Individuals.astype(np.float)
         return data_df
```

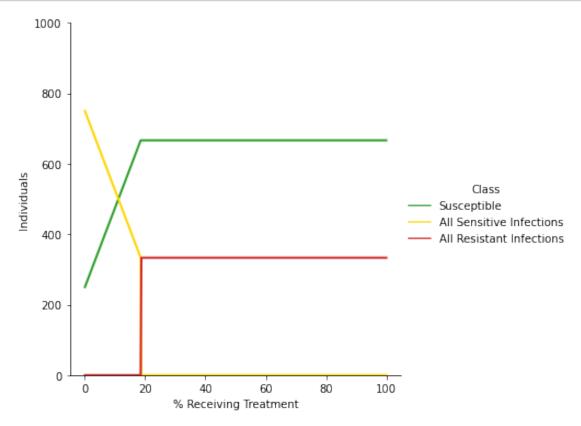
- 1 Figure 1 Finding the Equilibrium prevalences via deterministic simulation.
- 1.1 Singly infected models.
- 1.1.1 Figure 1a Exclusive Infection model Endemic Equiliruim

```
[6]: #Simulating the model
     data = run_model_for_epsilon_values(model=models_equil.AMR_exclus_inf_intia(),
                                         param_values=default_param_values,
      →epsilon_vals_as_percent=epsilon_vals_as_percent,
                                         time_range=time_range)
     data['Model'] = 'Exclusive Infection'
     deterministic_sims.append(data)
    % Receiving Treatment: 100%|
                                      | 401/401 [01:04<00:00, 6.25it/s]
[7]: # Creating figure of simulation
     fig = sns.relplot(
         data=data,
         x='% Receiving Treatment', y='Individuals',
         hue="Class", hue_order=classes_to_plot,
         kind="line", palette=palette, linewidth=2,
         aspect=1)
     fig.set(ylim=(0,N))
     plt.show()
```



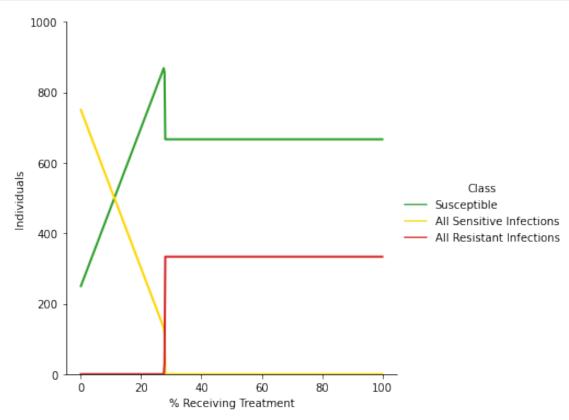
```
[9]: fig = sns.relplot(
    data=data,
    x='% Receiving Treatment', y='Individuals',
    hue="Class", hue_order=classes_to_plot,
    kind="line", palette=palette, linewidth=2,
    aspect=1)
```

```
fig.set(ylim=(0,N))
plt.show()
```



### 1.1.2 Figure 1b Replacement Infection model Endemic Equiliruim

```
aspect=1)
fig.set(ylim=(0,N))
plt.show()
```



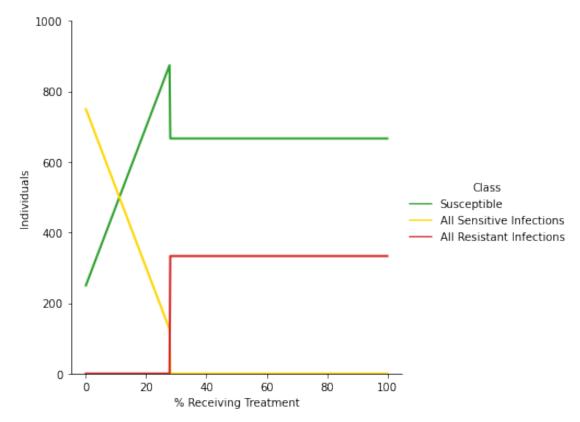
```
[12]: # Determining endemic equilibia prevelance through stable point analyses.
data = run_equilibria_func_for_epsilon_values(equilibria_func=models_equil.

→replace_inf_end_equil,

param_values=default_param_values,

→epsilon_vals_as_percent=epsilon_vals_as_percent)
```

```
fig = sns.relplot(
    data=data,
    x='% Receiving Treatment', y='Individuals',
    hue="Class", hue_order=classes_to_plot,
    kind="line", palette=palette, linewidth=2,
    aspect=1)
fig.set(ylim=(0,N))
plt.show()
```



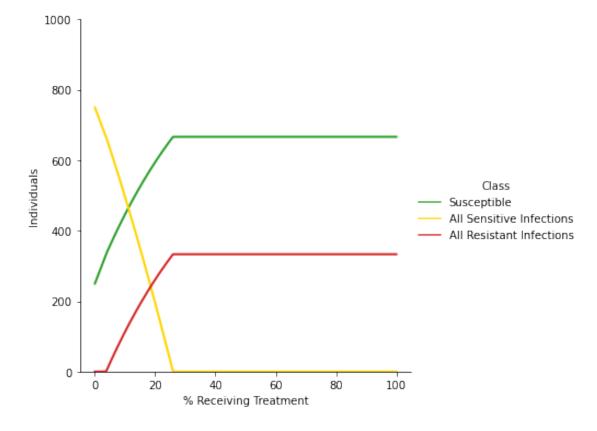
#### 1.2 Double Infection models.

### 1.2.1 Figure 1c Superinfection model

Note The equilibria for this model has not been derived.

```
data['Model'] = 'Superinfection'
deterministic_sims.append(data)
```

% Receiving Treatment: 100% | 401/401 [00:51<00:00, 7.80it/s]



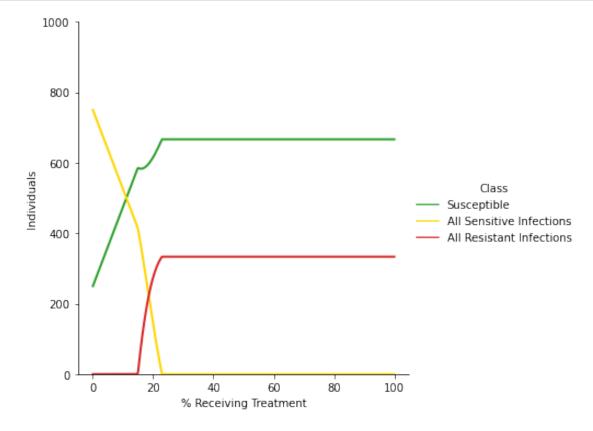
### 1.2.2 Figure 1d Full Coinfection model

```
→epsilon_vals_as_percent=epsilon_vals_as_percent,

time_range=time_range)

data['Model'] = 'Full Coinfection'
deterministic_sims.append(data)
```

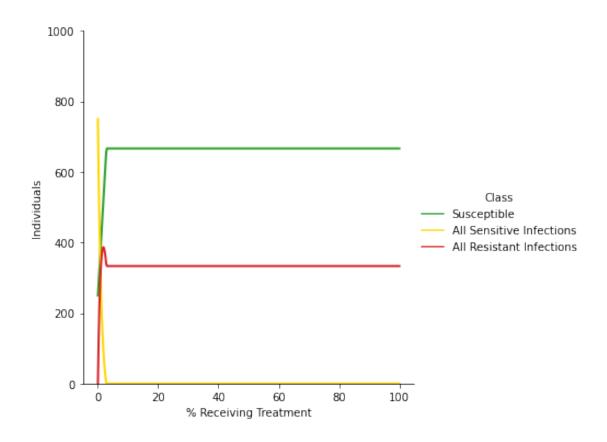
% Receiving Treatment: 100% | 401/401 [00:53<00:00, 7.52it/s]



### 1.3 Conversion models

### 1.3.1 Figure 1e Uni-direction Conversion model

```
[18]: #Simulating the model
      data = run_model_for_epsilon_values(model=models_equil.
       →AMR_bi_and_uni_conversion_inf_intia(),
                                           param_values={**default_param_values, 'rho':
       \rightarrow 0.5, 'phi':0},
      →epsilon_vals_as_percent=epsilon_vals_as_percent,
                                           time range=time range)
      data['Model'] = 'Unidirectional Conversion'
      deterministic_sims.append(data)
     % Receiving Treatment: 100%|
                                       | 401/401 [00:49<00:00, 8.07it/s]
[19]: # Creating figure of simulation
      fig = sns.relplot(
          data=data,
          x='% Receiving Treatment', y='Individuals',
          hue="Class", hue_order=classes_to_plot,
          kind="line", palette=palette, linewidth=2,
          aspect=1)
      fig.set(ylim=(0,N))
      plt.show()
```



```
[20]: # Determining endemic equilibria prevalence through stable point analyses.

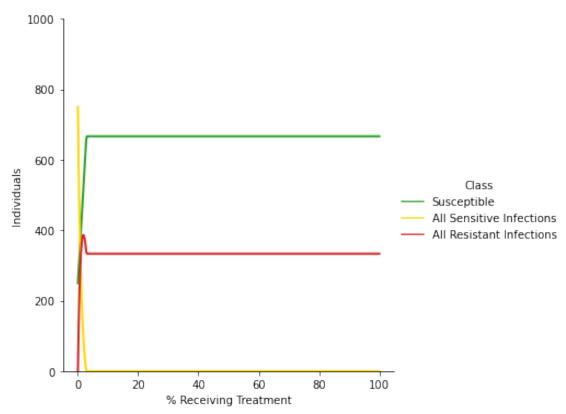
data = run_equilibria_func_for_epsilon_values(equilibria_func=models_equil.

→bi_directional_end_equil,

→param_values={**default_param_values, 'rho':0.5,'phi':0},

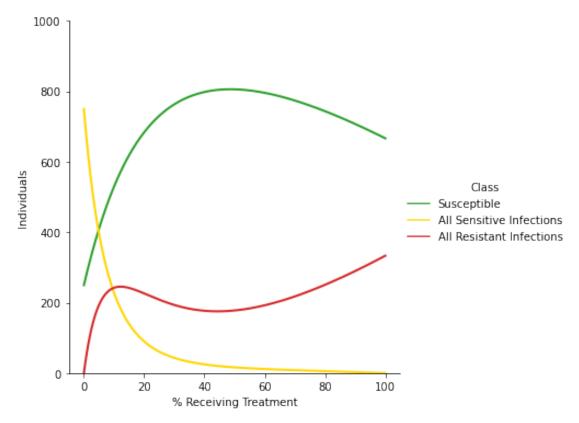
→epsilon_vals_as_percent=epsilon_vals_as_percent)
```

```
aspect=1)
fig.set(ylim=(0,N))
plt.show()
```



### 1.4 Figure 1f Bi-direction Conversion model

```
data=data,
    x='% Receiving Treatment', y='Individuals',
    hue="Class", hue_order=classes_to_plot,
    kind="line", palette=palette, linewidth=2,
    aspect=1)
fig.set(ylim=(0,N))
plt.show()
```



```
[24]: # Determining endemic equilibria prevalence through stable point analyses.
data = run_equilibria_func_for_epsilon_values(equilibria_func=models_equil.

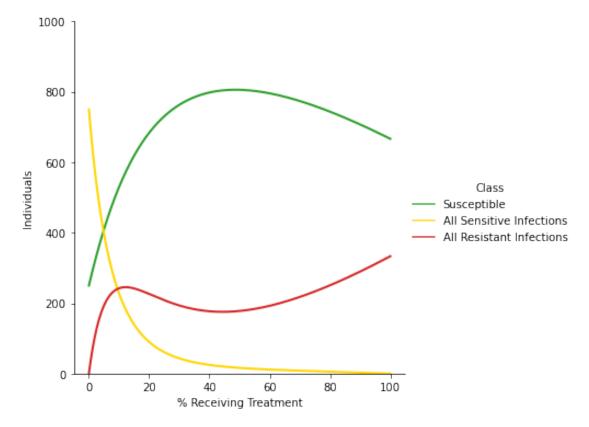
→bi_directional_end_equil,

→param_values={**default_param_values, 'rho':0.5,'phi':0.05},

→epsilon_vals_as_percent=epsilon_vals_as_percent)
```

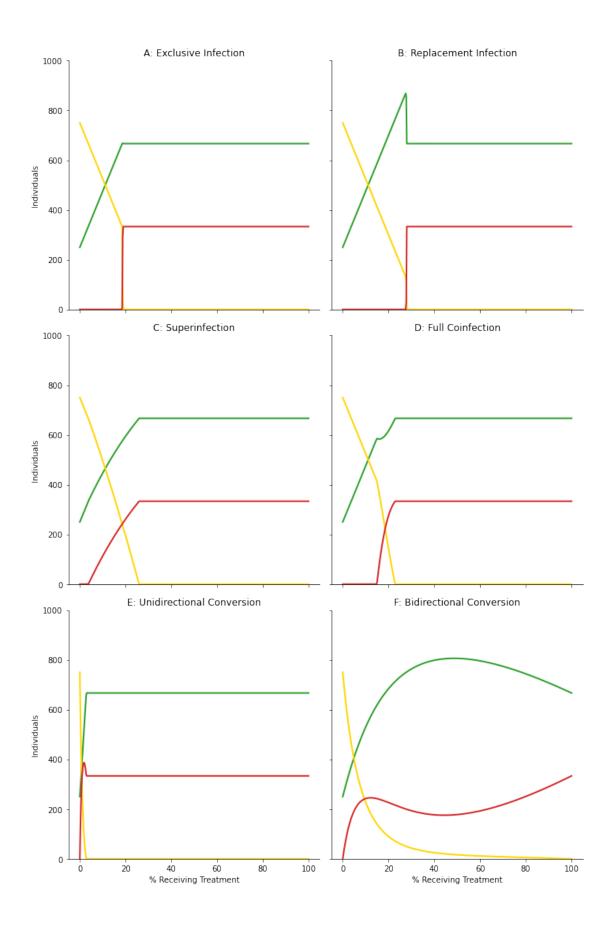
% Receiving Treatment: 100% | 401/401 [05:38<00:00, 1.19it/s] <ipython-input-5-cd4d980ac256>:25: DeprecationWarning: `np.float` is a deprecated alias for the builtin `float`. To silence this warning, use `float` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.float64` here. Deprecated in NumPy 1.20; for more details and guidance:

```
https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations
  data_df.Individuals = data_df.Individuals.astype(np.float)
```



## 2 Merging all simulations dataframes and creating a single figure.

```
fig = sns.relplot(
   data=full_data_df,
   x='% Receiving Treatment', y='Individuals',
   hue="Class", hue_order=classes_to_plot,
   col="Model", col_order=mod_order, col_wrap=2,
   kind="line", palette=palette,legend=False, linewidth=2,
   aspect=1)
# ADJUST ALL AXES TITLES
fig.set_titles('{col_name}')
alpha = list(string.ascii_uppercase)
axes = fig.axes.flatten()
for ax, letter in zip(axes, alpha[:len(axes)]):
   ttl = ax.get_title() # GET CURRENT TITLE
                                      # SET NEW TITLE
   ax.set_title(f"{letter}: {ttl}")
fig.set(ylim=(0,N))
plt.tight_layout() # automatically adjust spacing of axes
plt.savefig('Deterministic Simulations.png')
```



[27]: