#### Figures - CM AM

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To download and load the package simCAM (Simulations of Compartment and Agent-based Models), run

```
devtools::install_github("shannong19/simCAM")
library(simCAM)
```

You will need the following libraries for plotting:

```
library(ggplot2)
library(RColorBrewer)
library(reshape2)
library(gridExtra)
```

#### SIR-system

To run the stochastic SIR-CM, run the following commands. First run the deterministic SIR-CM to obtain the probabilities of transition for a given set of parameters.

Next, run the stochastic SIR-CM. We have set the number of simulations/runs equal to L=50. In the manuscript, we use 5000 runs.

```
L <- 50 # number of runs

## Set up of initial values
S <- matrix(0, nrow = T + 1, ncol = L)
I <- matrix(0, nrow = T + 1, ncol = L)
R <- matrix(0, nrow = T + 1, ncol = L)
cm_sim_list <- list(S = S, I = I, R = R)
cm_sim_list <- fill_inits_sir(init_vals, cm_sim_list)

## Run the simulation
cm_sim_list <- run_cm_sir(T, init_vals, probs, L, cm_sim_list)</pre>
```

Then, run the stochastic SIR-AM, also with L=50 simulations. The function summarize\_agents() puts the output agents into a format that is identical to stochastic SIR-CM output.

```
## Initialize
L <- 50
T <- 100
agents <- vector("list", L)
agents <- initialize_agents(T, N, init_vals, agents)

## Run the AM
out_agents <- run_am_sir(T, probs, L, agents)

## Summarize the agents
agents_sims <- summarize_agents(out_agents)
am_sim_list <- agents_sims</pre>
```

We then create our first set of figures. Figure (1) is the mean proportion of agents in each state for a given time for the SIR-system.

# Mean proportion in states

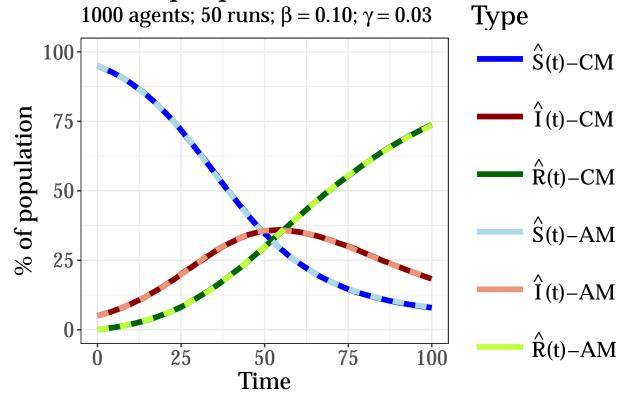


Figure (2) is the variance of of the agents in each state for a given time for the SIR-system.

```
## Variance
## Plotting the variance
am_var <- summary_sims(am_sim_list,
    fxn = rowVar,
var_names = c("S", "I", "R"))
cm_var <- summary_sims(cm_sim_list,
    fxn = rowVar,
    var_names = c("S", "I", "R"))
g <- plot_overlap(cm_var, am_var,
    summary_name = "Variance in states",
    ylim = c(0, max(max(cm_var), max(am_var))),
    plot_dash = FALSE,
    size = 2)
g</pre>
```

### Variance in states

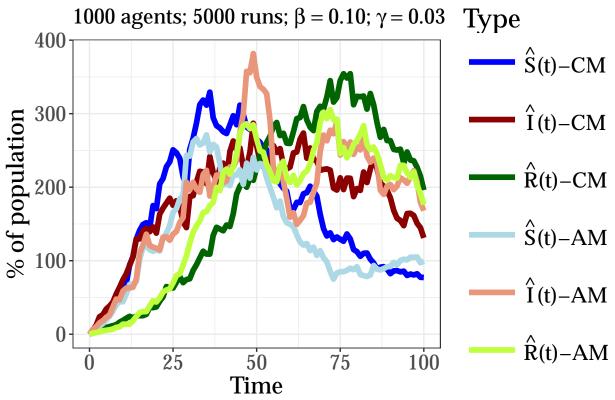
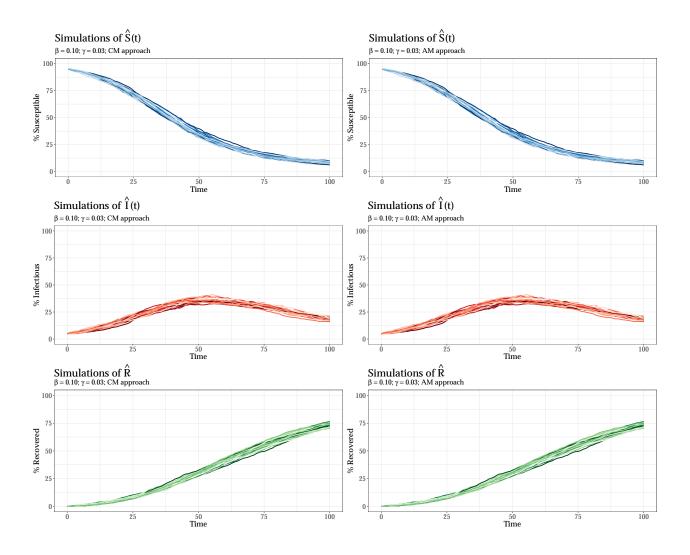


Figure (3) is the actual simulations (sample paths) for each state, for both the CM and AM.

```
cols <- c("Blues", "Reds", "Greens")
cm_titles <- c("Susceptible",
"Infectious",
"Recovered")
cm_symbols <- c("\\hat{S}(t)",
"\\hat{I}(t)",
"\\hat{R}")
g_list_cm <- lapply(1:length(cm_sim_list),
function(ind) {
plot_draws_sir(</pre>
```

```
cm_sim_list[[ind]],
beta,
gamma,
N,
L,
col = cols[ind],
cat_title = cm_titles[ind],
tex_symbol = cm_symbols[ind],
approach = "CM"
})
g_list_am <- lapply(1:length(am_sim_list),</pre>
function(ind) {
plot_draws_sir(
cm_sim_list[[ind]],
beta,
gamma,
N,
L,
col = cols[ind],
cat_title = cm_titles[ind],
tex_symbol = cm_symbols[ind],
approach = "AM"
)
})
## Plot on a grid
do.call("grid.arrange", c(g_list_cm, g_list_am, ncol=2, as.table = FALSE))
```



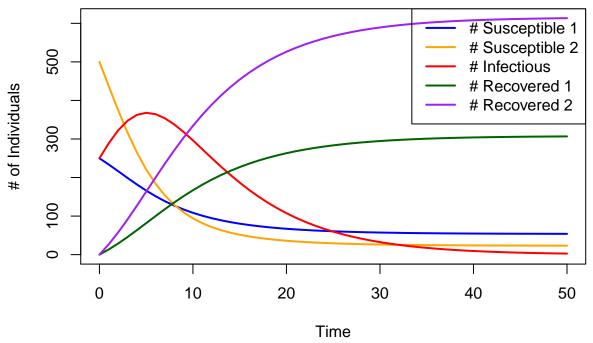
#### $S^2IR^2$ -system

We also have similar simulations for the  $S^2IR^2$ -system.

We first extract the probabilities from the deterministic  $S^2IR^2$ -CM with the following ste of initial parameters.

```
T <- 50
init_vals <- c(250, 500, 250, 0, 0)
N <- sum(init_vals)
beta1 <- .25
beta2 <- .5
gamma1 <- .05
gamma2 <- .1
step <- 1
inner_fxn <- SIR2_inner
results <- SIR2(T, init_vals, beta1, beta2, gamma1, gamma2)</pre>
```

S2IR2 curve beta1 = 0.25; beta2 = 0.5; gamma1 = 0.05; gamma2 = 0.1



```
## Extract the probabilities of transition
p <- extract_probs(results)</pre>
```

We then run the stochastic  $S^2IR^2$ -CM and stochastic  $S^2IR^2$ -AM.

```
## CM
L <- 50 # number of runs
S1 \leftarrow matrix(0, nrow = T + 1, ncol = L)
S2 \leftarrow matrix(0, nrow = T + 1, ncol = L)
I \leftarrow matrix(0, nrow = T + 1, ncol = L)
R1 \leftarrow matrix(0, nrow = T + 1, ncol = L)
R2 \leftarrow matrix(0, nrow = T + 1, ncol = L)
cm_sim_list <- list(</pre>
S1 = S1,
S2 = S2,
I = I,
R1 = R1,
R2 = R2
cm_sim_list <- fill_in_init_vals(init_vals, cm_sim_list)</pre>
cm_sim_list <- run_cm(T, init_vals, p, L = L, cm_sim_list)</pre>
## AM
L <- 50
T < -50
agents <- vector("list", L)</pre>
agents <- initialize_agents(T, N, init_vals, agents)</pre>
out_agents <- run_am(T, p, L, agents)</pre>
```

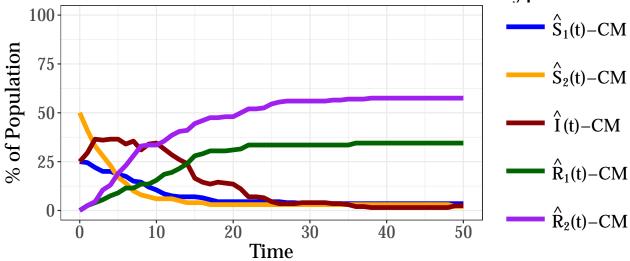
```
## Put agents in same form as CM output
agents_sims <- summarize_agents(out_agents, n_states = 5)
am_sim_list <- agents_sims</pre>
```

We then make our analogous plots of the first three figures, Figures (5)-(7).

```
## Mean
## CM
cm_mean <- summary_sims(cm_sim_list)
g_avg_cm <-
plot_summary(
cm_mean,
L = L,
beta1 = beta1,
beta2 = beta2,
gamma1 = gamma1,
gamma2 = gamma2)
)
g_avg_cm</pre>
```

## Mean Proportion of State Values -- CM

1000 agents; 50 runs;  $\beta_1 = 0.25$ ;  $\beta_2 = 0.50$ ;  $\gamma_1 = 0.05$ ;  $\gamma_2 = 0.1$  **Pype** 

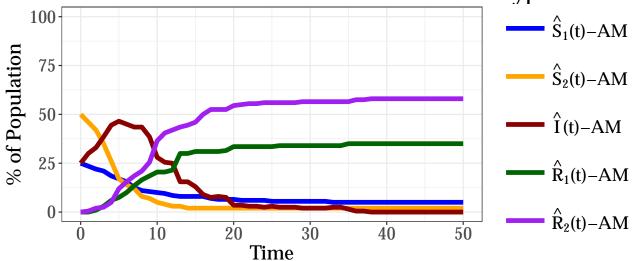


```
## AM
am_mean <- summary_sims(am_sim_list)

g_avg_am <- plot_summary(
am_mean,
approach = "AM",
L = L,
beta1 = beta1,
gamma1 = gamma1,
beta2 = beta2,
gamma2 = gamma2
)
g_avg_am</pre>
```

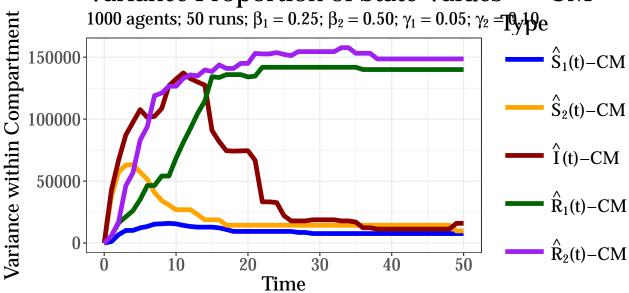
## Mean Proportion of State Values -- AM

1000 agents; 50 runs;  $\beta_1 = 0.25$ ;  $\beta_2 = 0.50$ ;  $\gamma_1 = 0.05$ ;  $\gamma_2 = 0.1$  pype



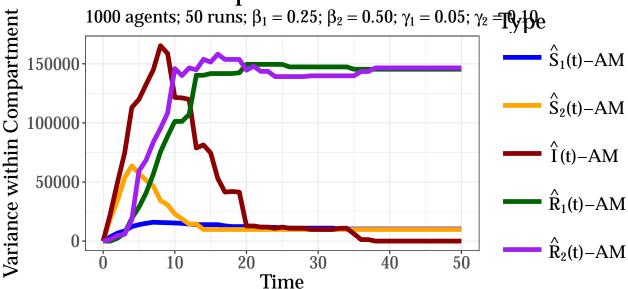
```
## CM
cm_var <- summary_sims(cm_sim_list, fxn = rowVar)
g_var_cm <-
plot_summary(
cm_var,
sum_name = "Variance Proportion",
ylab = "Variance within Compartment",
L = L,
beta1 = beta1,
gamma1 = gamma1,
beta2 = beta2,
gamma2 = gamma2
)
g_var_cm</pre>
```

## Variance Proportion of State Values -- CM



```
## AM
am_var <- summary_sims(am_sim_list, fxn = rowVar)
g_var_am <-
plot_summary(
am_var,
sum_name = "Variance Proportion",
ylab = "Variance within Compartment",
approach = "AM",
L = L,
beta1 = beta1,
gamma1 = gamma1,
beta2 = beta2,
gamma2 = gamma2
)
g_var_am</pre>
```

## Variance Proportion of State Values -- AM



```
## Plotting all the draws
cols <- c("Blues", "Oranges", "Reds", "Greens", "Purples")</pre>
cm_titles <- c("Susceptible 1",</pre>
"Susceptible 2",
"Infectious",
"Recovered 1",
"Recovered 2")
cm_symbols <- c("\hat{S}_1(t)",
\hfill

"\\hat{S}_2(t)",
\hfill \{I\}(t),
"\\hat{R}_1(t)",
"\\hat{R}_2(t)")
g_list_cm <- lapply(1:length(cm_sim_list),</pre>
                     function(ind) {
                     plot_draws_sir(
                     cm_sim_list[[ind]],
                     beta,
                     gamma,
                     N,
                     L,
                     col = cols[ind],
                     cat_title = cm_titles[ind],
                     tex_symbol = cm_symbols[ind],
                     approach = "CM"
                     )
                     })
g_list_am <- lapply(1:length(am_sim_list),</pre>
function(ind) {
plot_draws_s2ir2(
am_sim_list[[ind]],
beta1,
beta2,
```

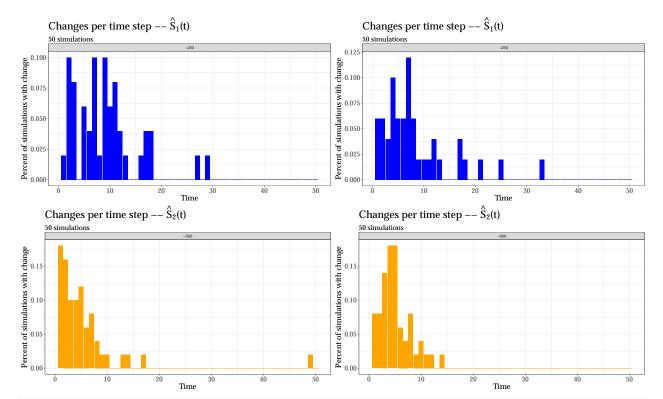
```
gamma1,
gamma2,
N,
L,
col = cols[ind],
cat_title = cm_titles[ind],
tex_symbol = cm_symbols[ind],
approach = "AM"
)
})
## Plotting
do.call("grid.arrange", c(g_list_cm, g_list_am, ncol=2, as.table = FALSE))
       Simulations of \hat{S}_1(t)
                                                                                                                   Simulations of \hat{S}_1(t)
                                                                                                                  \beta_1 = 0.25; \, \beta_2 = 0.50; \, \gamma_1 = 0.05; \, \gamma_2 = 0.10; \, AM \; approach
       \beta = 0.10; \gamma = 0.03; CM approach
% Susceptible 1
                                                                                                          20
                                                     Time
                                                                                                                                                                 Time
                                                                                                                   Simulations of \hat{S}_2(t)
       Simulations of \hat{S}_2(t)
       \beta=0.10;\,\gamma=0.03;\,CM approach
                                                                                                                   \beta_1 = 0.25; \, \beta_2 = 0.50; \, \gamma_1 = 0.05; \, \gamma_2 = 0.10; \, AM \; approach
% Susceptible 2
                                                                                                          % Susceptible 2
                                                                                                               75
                                                                                                               50
25
       Simulations of \hat{I}(t)
                                                                                                                  Simulations of \hat{I}(t)
       \beta = 0.10; \gamma = 0.03; CM approach
                                                                                                                  \beta_1 = 0.25; \, \beta_2 = 0.50; \, \gamma_1 = 0.05; \, \gamma_2 = 0.10; \, AM \; approach
100 75 50 25 0
                                                                                                          % Infectious
                                                                                                               75
                                                                                                               50
                                                                                                               25
       Simulations of \hat{R}_1(t)
                                                                                                                  Simulations of \hat{R}_1(t)
       \beta=0.10;\,\gamma=0.03;\,CM approach
                                                                                                                  \beta_1 = 0.25; \, \beta_2 = 0.50; \, \gamma_1 = 0.05; \, \gamma_2 = 0.10; \, AM \,\, approach
Recovered
                                                                                                               75
                                                                                                               50-
                                                                                                               25
                                                     Time
                                                                                                                                                                 Time
       Simulations of \hat{R}_2(t)
                                                                                                                  Simulations of \hat{R}_2(t)
                                                                                                                  \beta_1 = 0.25; \, \beta_2 = 0.50; \, \gamma_1 = 0.05; \, \gamma_2 = 0.10; \, AM \; approach \,
       \beta = 0.10; \gamma = 0.03; CM approach
Recovered 2
0 % Recovered 2
0 % Recovered 3
                                                                                                           % Recovered 2
                                                                                                               75 ·
50 ·
                                                                                                               25
                                                     Time
                                                                                                                                                                 Time
```

#### Time Distributions

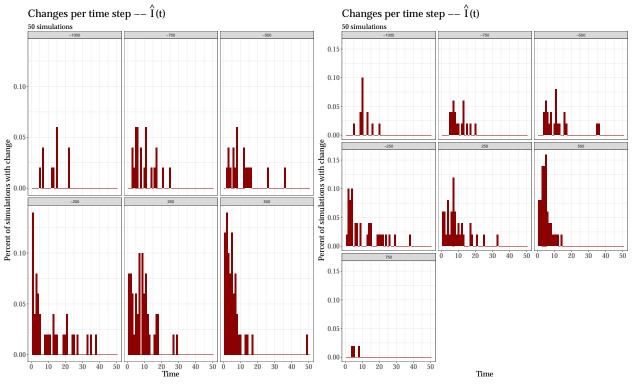
Figures (8)-(11) pertain to transitions from compartments over time. These graphs are produced with the following.

```
cols <- c("blue", "orange", "darkred", "darkgreen", "purple")
cm_titles <- c("Susceptible 1",</pre>
```

```
"Susceptible 2",
"Infectious",
"Recovered 1",
"Recovered 2")
cm_symbols <- c("\hat{S}_1(t)",
"\\hat{S}_2(t)",
"\\hat{I}(t)",
"\\hat{R}_1(t)",
"\\hat{R}_2(t)")
g_list_cm_times <- lapply(1:length(cm_sim_list),</pre>
function(ind) {
plot_time_dist(
cm_sim_list[[ind]],
beta1,
beta2,
gamma1,
gamma2,
N,
L,
col = cols[ind],
cat_title = cm_titles[ind],
tex_symbol = cm_symbols[ind]
)
})
g_list_am_times <- lapply(1:length(am_sim_list),</pre>
function(ind) {
plot_time_dist(
am_sim_list[[ind]],
beta1,
beta2,
gamma1,
gamma2,
N,
L,
col = cols[ind],
cat_title = cm_titles[ind],
tex_symbol = cm_symbols[ind]
})
## Arrange
grid.arrange(g_list_cm_times[[1]],
g_list_am_times[[1]],
g_list_cm_times[[2]],
g_list_am_times[[2]],
ncol = 2)
```

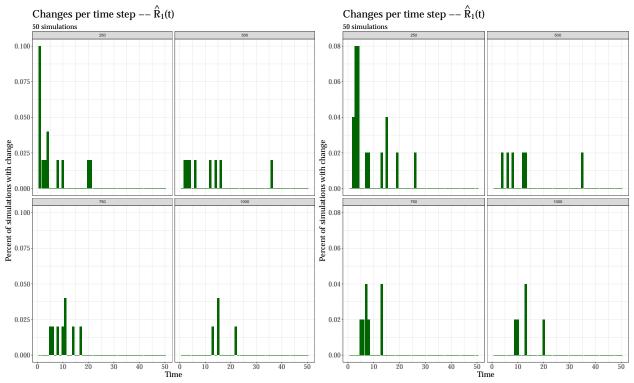


## I
grid.arrange(g\_list\_cm\_times[[3]], g\_list\_am\_times[[3]],
ncol = 2)



## R1
grid.arrange(g\_list\_cm\_times[[4]], g\_list\_am\_times[[4]],

#### ncol = 2)



## R2
grid.arrange(g\_list\_cm\_times[[5]], g\_list\_am\_times[[5]],
ncol = 2)

