

11. Using R to Support Simulation Input and Output Analysis


A Case Study with System Dynamics

Using the tidyverse to support simulation

- ▶ Tidying input data
- ▶ Analysing simulation output
- ▶ Running sensitivity analysis

NOTES AND INSIGHTS

Input and output data analysis for system dynamics modelling using the tidyverse libraries of R

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(1) Tidying input data

Table 2. Time series influenza data from the 1957 pandemic (U.K. data)

Week	Young	Child	Adult	Elderly
1	0	0	1	1
2	0	2	6	1
3	0	2	4	2
4	23	73	63	11
5	63	208	173	41
6	73	207	171	27
7	66	150	143	7
8	26	40	87	29
9	17	18	33	12
10	3	4	13	6
11	2	6	16	5
12	1	6	11	3
13	0	1	6	5
14	0	2	2	2
15	0	1	3	0
16	0	1	4	6
17	0	1	3	0
18	2	1	7	1
19	1	1	6	2

Using `readr` to access data

```
inc <- read_csv("../..//11 simulation/code/sdr_paper1/data/1
```

```
## Parsed with column specification:
```

```
## cols(
```

```
##   Week = col_double(),
```

```
##   Young = col_double(),
```

```
##   Child = col_double(),
```

```
##   Adult = col_double(),
```

```
##   Elderly = col_double()
```

```
## )
```

```
slice(inc,1:2)
```

```
## # A tibble: 2 x 5
```

```
##   Week Young Child Adult Elderly
```

```
##   <dbl> <dbl> <dbl> <dbl>   <dbl>
```

```
## 1     1     0     0     1     1
```

```
## 2     2     0     2     6     1
```

Convert to Tidy Data

```
t_inc <- gather(inc,Cohort,Incidence,Young:Elderly)
slice(t_inc,1:8)
```

```
## # A tibble: 8 x 3
##   Week Cohort Incidence
##   <dbl> <chr>      <dbl>
## 1     1  1 Young          0
## 2     2  2 Young          0
## 3     3  3 Young          0
## 4     4  4 Young         23
## 5     5  5 Young         63
## 6     6  6 Young         73
## 7     7  7 Young         66
## 8     8  8 Young         26
```

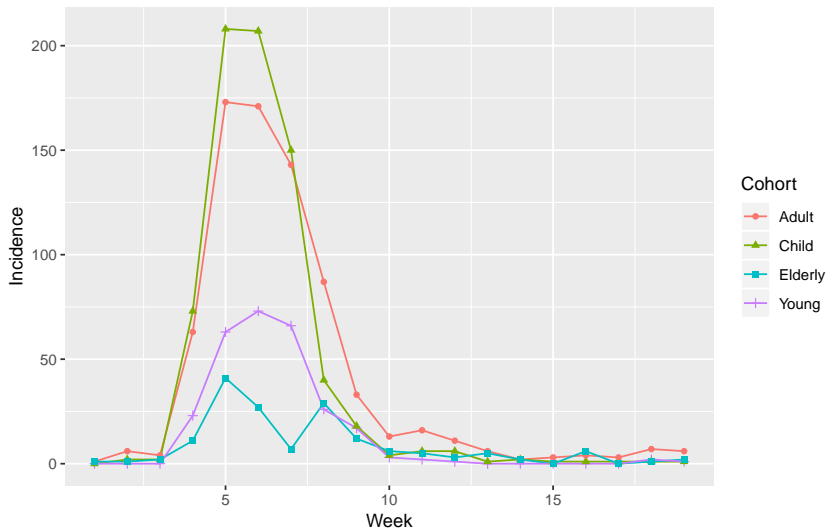
Summarise Data

```
wk_tot <- t_inc %>% group_by(Week) %>%  
  summarise(Incidence=sum(Incidence)) %>%  
  arrange(desc(Incidence))  
slice(wk_tot,1:6)
```

```
## # A tibble: 6 x 2  
##   Week Incidence  
##   <dbl>     <dbl>  
## 1     5       485  
## 2     6       478  
## 3     7       366  
## 4     8       182  
## 5     4       170  
## 6     9        80
```

Plot tidy data

```
ggplot(t_inc, aes(x=Week, y=Incidence, color=Cohort,  
                  shape=Cohort)) +  
  geom_line() + geom_point()
```



Descriptive Statistics

```
t_coh <- t_inc %>%  
  group_by(Cohort) %>%  
  summarise(TotalInfected=sum(Incidence),  
            PeakValue=max(Incidence),  
            PeakWeek=Week[which(Incidence==max(Incidence))],  
            AvrValue=mean(Incidence),  
            SD=sd(Incidence))  
t_coh
```

```
## # A tibble: 4 x 6
```

```
##   Cohort   TotalInfected PeakValue PeakWeek AvrValue      S  
##   <chr>         <dbl>      <dbl>    <dbl>   <dbl> <dbl>  
## 1 Adult           752         173        5    39.6  59  
## 2 Child           724         208        5    38.1  70  
## 3 Elderly         161          41        5     8.47 11  
## 4 Young          277          73        6    14.6  24
```


(2) Analysing Simulation Output

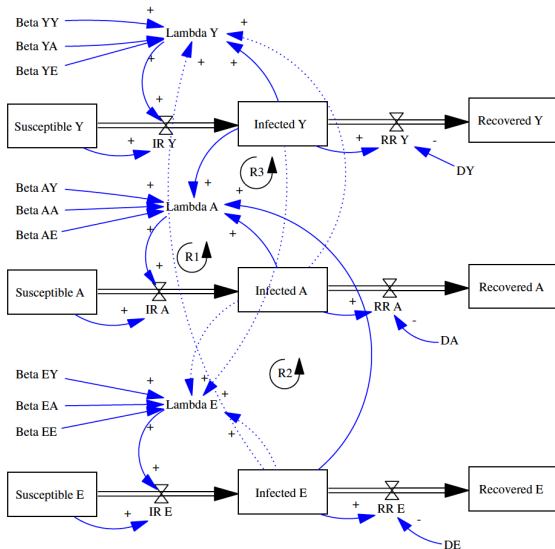


Fig. 3. The SIR Model for three cohorts [Color figure can be viewed at wileyonlinelibrary.com]

Simulation results - many columns

##	[1]	"Time"	"Beta AA"	"Beta AE"
##	[4]	"Beta AY"	"Beta EA"	"Beta EE"
##	[7]	"Beta EY"	"Beta YA"	"Beta YE"
##	[10]	"Beta YY"	"CE AA"	"CE AE"
##	[13]	"CE AY"	"CE EA"	"CE EE"
##	[16]	"CE EY"	"CE YA"	"CE YE"
##	[19]	"CE YY"	"DA"	"DE"
##	[22]	"DY"	"Infected A"	"Infected E"
##	[25]	"Infected Y"	"IR A"	"IR E"
##	[28]	"IR Y"	"Lambda A"	"Lambda E"
##	[31]	"Lambda Y"	"Pop A"	"Pop E"
##	[34]	"Pop Y"	"Prop A Infected"	"Prop E Infected"
##	[37]	"Prop Y Infected"	"Recovered A"	"Recovered E"
##	[40]	"Recovered Y"	"RR A"	"RR E"
##	[43]	"RR Y"	"Susceptible A"	"Susceptible E"
##	[46]	"Susceptible Y"	"Total Population"	

Selecting the stocks

```
out <- res %>%  
  select(Time, starts_with("Susceptible"),  
         starts_with("Infected"),  
         starts_with("Recovered"))  
glimpse(out)
```

```
## Observations: 161
```

```
## Variables: 10
```

```
## $ Time          <dbl> 0.000, 0.125, 0.250, 0.375, 0.500
```

```
## $ `Susceptible A` <dbl> 50000, 50000, 50000, 50000, 50000
```

```
## $ `Susceptible E` <dbl> 25000, 25000, 25000, 25000, 25000
```

```
## $ `Susceptible Y` <dbl> 25000, 25000, 25000, 25000, 25000
```

```
## $ `Infected A`    <dbl> 0.00000, 0.00000, 0.01562, 0.05469
```

```
## $ `Infected E`    <dbl> 0.0000, 0.1250, 0.2891, 0.5083,
```

```
## $ `Infected Y`    <dbl> 1.000, 1.312, 1.738, 2.321, 3.125
```

```
## $ `Recovered A`   <dbl> 0.00000, 0.00000, 0.00000, 0.00000
```

```
## $ `Recovered E`   <dbl> 0.00000, 0.00000, 0.00781, 0.02500
```

```
## $ `Recovered Y`   <dbl> 0.0000, 0.0625, 0.1445, 0.2532,
```

Convert to tidy format

```
out_td <- out %>%  
  gather(key=Variable,value = Amount,  
         `Susceptible A`:`Recovered Y`)  
slice(out_td,1:5)
```

```
## # A tibble: 5 x 3  
##   Time Variable      Amount  
##   <dbl> <chr>      <dbl>  
## 1 0      Susceptible A  50000  
## 2 0.125 Susceptible A  50000  
## 3 0.25  Susceptible A  50000  
## 4 0.375 Susceptible A  50000  
## 5 0.5   Susceptible A  50000
```

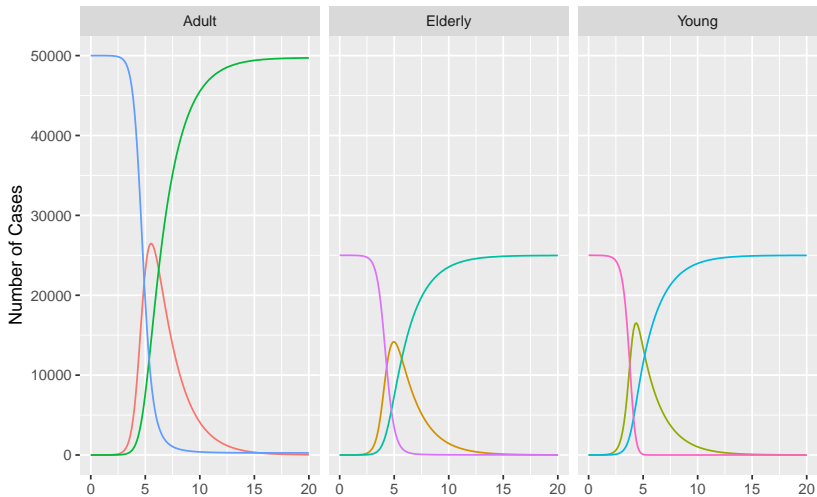
Add cohort and stock information

```
new_td <- out_td %>%  
  mutate(Cohort=case_when(  
    grepl("A$",Variable) ~ "Adult",  
    grepl("E$",Variable) ~ "Elderly",  
    grepl("Y$",Variable) ~ "Young"),  
    Class=case_when(  
    grepl("^S",Variable) ~ "Susceptible",  
    grepl("^I",Variable) ~ "Infected",  
    grepl("^R",Variable) ~ "Recovered"))  
slice(new_td,1:3)
```

```
## # A tibble: 3 x 5  
##   Time Variable      Amount Cohort Class  
##   <dbl> <chr>          <dbl> <chr> <chr>  
## 1 0      Susceptible A  50000 Adult  Susceptible  
## 2 0.125 Susceptible A  50000 Adult  Susceptible  
## 3 0.25  Susceptible A  50000 Adult  Susceptible
```

Display chart

```
ggplot(new_td) +  
  geom_path(aes(x=Time,y=Amount,colour=Variable))+  
  ylab("Number of Cases")+  
  facet_wrap(~Cohort)+guides(colour=F)
```



(3) Exploring Sensitivity Data

```
d <- read_tsv("../..//11 simulation/code/sdr_paper1/data/Ser  
dim(d)
```

```
## [1] 200 244
```

```
d[1:3,1:5]
```

```
## # A tibble: 3 x 5
```

	Simulation	R0	VF	`T1 Infected`	`T2 Infected`
##	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
## 1	1	2.76	0.0263	1	1.11
## 2	2	2.66	0.0739	1	1.10
## 3	3	4.06	0.159	1	1.19

Convert to Tidy Data

```
START_TIME <- 0
DT <- 0.125

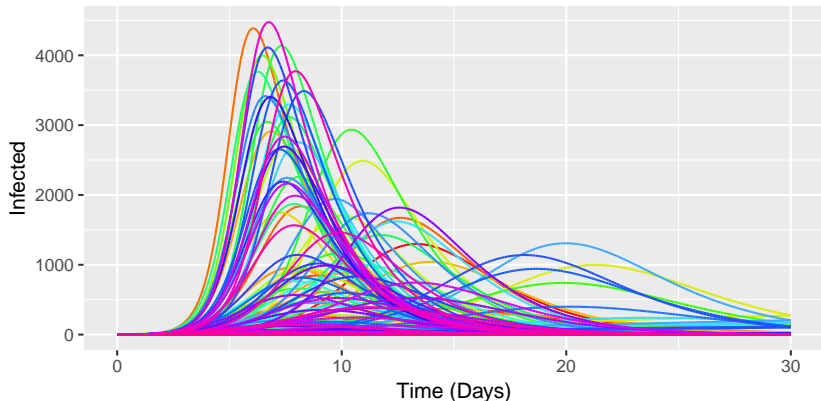
td <- gather(d, TimeVariable, Value, -(Simulation:VF)) %>%
  mutate(TSeq=parse_integer(
    str_extract(TimeVariable, "\\d+"))) %>%
  mutate(SimTime=START_TIME+(TSeq-1)*DT) %>%
  separate(TimeVariable, into = c("T", "Variable")) %>%
  select(Simulation, SimTime, R0, VF, Variable, Value) %>%
  arrange(Simulation, SimTime)

slice(td, 1:2)
```

```
## # A tibble: 2 x 6
##   Simulation SimTime    R0      VF Variable Value
##       <dbl>   <dbl> <dbl>   <dbl> <chr>   <dbl>
## 1         1     0     2.76 0.0263 Infected 1
## 2         1 0.125     2.76 0.0263 Infected 1.11
```


Display simulation traces

```
ggplot(td,aes(x=SimTime,y=Value,color=Simulation)) +  
  geom_path() +  
  scale_colour_gradientn(colours=rainbow(10)) +  
  ylab("Infected") + xlab("Time (Days)") +  
  guides(color=FALSE)
```



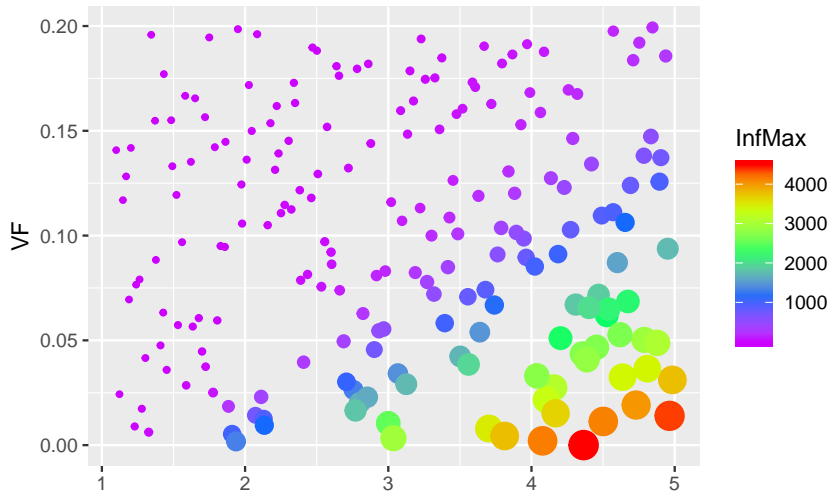
Calculate Summary Data

```
i_td <- td %>% group_by(Simulation) %>%  
  summarise(InfMax=max(Value),  
            R0=R0[1],  
            VF=VF[1])  
  
slice(i_td,1:5)
```

```
## # A tibble: 5 x 4  
##   Simulation InfMax    R0    VF  
##   <dbl>    <dbl> <dbl> <dbl>  
## 1         1 1298.   2.76 0.0263  
## 2         2  78.9   2.66 0.0739  
## 3         3  127.   4.06 0.159  
## 4         4  367.   2.69 0.0496  
## 5         5  14.0   3.23 0.194
```

Explore Parameter Space

```
ggplot(data=i_td,aes(x=R0,y=VF,size=InfMax,colour=InfMax))  
  geom_point() +  
  scale_colour_gradientn(colours=rev(rainbow(5)))+  
  guides(size=F)
```



Conclusion

Conclusion

While R is primarily viewed as a toolset to support data scientists, innovative new libraries such as the `tidyverse` can be leveraged to support the system dynamics model-building process. This paper has shown how time series data can be accessed and manipulated, and how the entire model output from a simulation run can be processed for informative summaries and for data visualisation. A further application of the `tidyverse` is to support the process of analysing large datasets produced through sensitivity analysis of system dynamics models.

Test Slide with Plot

