Package 'hte'

March 13, 2024

R topics documented:

hte-	package	Но	spi	tal	tra	ıns	mi.	ssi	on	e.	sti	m	ati	on	ı														
Index																													19
	twostate			•		•		•	•	•	•		•	•	•	•	•	•	 •	•	•	•	•	•	•	•	 •	•	18
	transmission																												
	stobfun																												
	simuldat																												
	independent																												7
	fake_data																												4
	coal_last																												3
	Bernoulli_filter																												2
	hte-package																												1

Description

The **hte** package ...

Author(s)

Aaron A. King

2 Bernoulli_filter

Description

Filter the individual patient data with given forces of infection, recovery rates, and test characteris-

Usage

```
Bernoulli_filter(data, lambda, gamma, theta)
Bfilter(data, theta)
```

Arguments

data data set

lambda unit-specific force of infection gamma unit-specific recovery rate

theta list of parameters

Details

Bernoulli_filter runs a Bernoulli filter, updating the expected prevalence. It returns the log likelihood, occupancy, and expected prevalences.

Bfilter uses an alternative algorithm.

Value

Bernoulli_filter returns a tibble containing the expected prevalences (for both isolated and un-isolated patients), unit occupancy, force of infection, and conditional log likelihood for each unit at each event time. The sum of the log likelihood column (logLik) is the log likelihood of the data

Bfilter returns the log likelihood of the furnished data.

```
library(tidyverse)
set.seed(626292345)

fake_data |>
    arrange(patient, time) -> dat

theta <- list(
    lambda=c(out=0, A=0.01, B=0.1, C=0.2, D=0.3, E=0.5),
    gamma=c(out=0.1, A=0.1, B=0.1, C=0.1, D=0.1, E=0.1),</pre>
```

coal_last 3

```
p0=0.2,
  isol_factor=0.1,
  alpha=0.05,
  beta=0.2,
  eta = 0.5
)
dat |>
  Bfilter(theta) -> ll1
sum(111)
dat |>
  Bernoulli_filter(
    lambda=theta$lambda,
   gamma=theta$gamma,
   theta
  ) -> f
f |> filter(logLik!=0) |> pull(logLik) -> 112
sum(112)
f
f |>
  select(unit,time,prev_i,prev_u) |>
  pivot_longer(c(prev_i,prev_u)) |>
  group_by(unit) |>
  ## prevalence is not estimated outside the hospital
  filter(!all(is.na(value))) |>
  ungroup() |>
  ggplot(aes(x=time,color=name,y=value))+
  geom_line(alpha=0.5)+
  scale_color_manual(values=c(prev_i="blue",prev_u="red"))+
  facet_grid(unit~.,labeller=label_both)+
  labs(y="prevalence")+
  theme_bw()
```

coal_last

Coalesce with last

Description

Fills NA with previous non-NA.

Usage

```
coal_last(x)
```

Arguments

Х

vector

fake_data

fake_data

Fake hospital movement, testing, and isolation data

Description

A simulated outbreak

Details

Data are generated using the default settings of simuldat.

The data were generated by:

```
set.seed(339613584)
simuldat(verbose=TRUE) -> fake_data
save(fake_data,file="fake_data.rda",compress="xz")
```

See Also

More on simulated data: simuldat()

```
library(tidyverse)
library(lubridate)
## Examine the data:
fake_data
## Verify certain conditions hold:
stopifnot(
  `admission condition violation`=fake_data |>
    group_by(patient,visit) |> slice_head() |>
    filter(event!="admit") |> nrow()==0,
  `discharge condition violation`=fake_data |>
    group_by(patient,visit) |> slice_tail() |>
    filter(event!="discharge", event!="stop") |> nrow()==0,
  `unit violation`=fake_data |> filter(is.na(unit)) |> nrow()==0,
  `event violation`=fake_data |> filter(is.na(event)) |> nrow()==0
)
fake_data |>
  mutate(
   time=as.numeric(
      as.duration(
        interval(date, start="2000-01-01T00:00:00+0000")
      units="day"
  ) -> fake_data
```

fake_data 5

```
fake_data |>
 group_by(patient, visit) |>
 summarize(dur=max(time)-min(time)) |>
 ungroup() |>
 group_by(patient) |>
 summarize(dur=sum(dur)) |>
 ggplot(aes(x=log10(dur)))+
 geom_histogram(bins=20)+
 labs(title="total duration of hospitalization")+
 theme_bw()
fake_data |>
 group_by(patient,visit) |>
 summarize(dur=max(time)-min(time)) |>
 ungroup() |>
 ggplot(aes(x=log10(dur)))+
 geom_histogram(bins=40)+
 labs(title="duration of hospital visit")+
 theme_bw()
fake_data |>
 filter(
    event!="test",
    event!="isolate",
   event!="release",
    event!="stop"
 group_by(patient, visit) |>
 arrange(time) |>
 mutate(dur=lead(time)-time) |>
 ungroup() |>
 filter(unit!="out",!is.na(dur)) |>
 ggplot(aes(x=log10(dur),fill=unit,group=unit))+
 geom_histogram(aes(y=after_stat(density)),bins=40)+
 facet_grid(unit~.,scales="free_y")+
 labs(title="duration of stay by unit")+
 theme_bw()
fake_data |>
 group_by(patient) |>
 summarize(ntest=sum(event=="test")) |>
 ungroup() |>
 ggplot(aes(x=ntest))+
 geom_histogram(binwidth=1,center=0)+
 labs(title="number of tests per patient")+
 theme_bw()
fake_data |>
 arrange(time) |>
 mutate(
   dn=case_when(
     event=="admit"~1L,
```

6 fake_data

```
event=="discharge"~-1L,
     TRUE~0L
   ),
   occ=cumsum(dn)
 ) |>
 ggplot(aes(x=date,y=occ))+
 geom_step()+
 labs(title="hospital occupancy")+
 theme_bw()
fake_data |>
 arrange(time) |>
 select(date,test.result=result,isol,infected) |>
 pivot_longer(c(test.result,isol,infected)) |>
 filter(!is.na(value)) |>
 ggplot(aes(x=date,y=value,color=name))+
 geom_point()+
 geom_smooth()+
 guides(color="none")+
 labs(
    title="infection and isolation status, test results",
   y=""
 )+
 facet_grid(name~.)+
 theme_bw()
fake_data |>
 filter(event=="test") |>
 mutate(
   interval=cut(time,breaks=72,ordered=TRUE)
 ) |>
 select(interval,time,infected,isol,result) |>
 pivot_longer(c(infected,isol,result)) |>
 group_by(name,interval) |>
 summarize(
   time=mean(time),
   prev=mean(value),
   n=n()
 ) |>
 ungroup() |>
 ggplot(aes(x=time,y=prev,group=name,fill=name))+
 geom_col(position="dodge")+
 labs(title="infection, isolation, and detection through time")+
 theme_bw()+
 theme(axis.text.x=element_text(angle=90))
fake_data |>
 filter(event=="test") |>
 select(infected,result) |>
 count(infected,result) |>
 group_by(infected) |>
 mutate(prob=n/sum(n)) |>
 ungroup()
```

independent 7

ependent Independent infection model

Description

Model under which each patient's trajectory of infection is independent, conditional on the force of infection.

Usage

```
indep_homog_filter(params, data)
indep_homog_objfun(params, data, est = character(0))
indep_unit_spec_filter(params, data)
indep_unit_spec_objfun(params, data, est = character(0))
```

Arguments

params named vector of parameters

data patient movement, isolation, and testing data

est names of parameters to estimate

Details

Parameters in the model include:

lambda force of infection. In the homogeneous model, this is constant across units in the hospital; in the unit-specific model, there is one value of lambda for each unit.

lambda.out force of infection outside the hospital

gamma recovery rate

alpha,beta false positive and negative probabilities

p0 probability of infection on admission

isol_factor multiplicative effect of contact isolation on susceptibility

eta additional information on infection status

indep_homog_filter runs a Bernoulli filter for the independent model with a global lambda and gamma.

indep_homog_objfun is a stateful objective function for the independent model with a global lambda and gamma.

indep_unit_spec_filter runs a Bernoulli filter for the independent model with a unit-specific lambda.

indep_unit_spec_objfun is a stateful objective function for the independent model with unit-specific lambda.

8 independent

See Also

More on the independent infection model: simuldat()

More on stateful objective functions: stobfun(), transmission

```
library(tidyverse)
library(lubridate)
library(pomp)
library(hte)
set.seed(626292345)
## Select some data:
fake_data |>
  filter(
   event!="test" | time < 250,
    time < 1000
  ) |>
  arrange(patient,time) -> dat
## Create an objective function:
indep_homog_objfun(
  params=c(
   lambda.out=0.05,lambda=0.1,gamma=0.01,p0=0.1,
    isol_factor=0.2,alpha=0.02,beta=0.1,eta = 0.5
  ),
  est=c("lambda", "gamma"),
  data=dat
) -> f
## Fit the model:
optim(
  par=log(c(0.1,0.01)),
  control=list(reltol=1e-3)
) -> out
f(out$par)
coef(f)
## Examine the filter results:
indep_homog_filter(params=coef(f),data=dat) -> ff
ff |> filter(logLik!=0)
## Construct a slice:
indep_homog_objfun(
  params=coef(f),
  est=c("lambda"),
  data=dat
) -> g
log.lambda \leftarrow seq(log(0.001), to=log(1), length=10)
```

```
plot(exp(log.lambda), sapply(log.lambda,g), log='x')
## Construct an objective function for the unit-specific model:
theta <- coef(f)
indep_unit_spec_objfun(
 params=c(
   lambda=setNames(rep.int(theta["lambda"],6),unique(dat$unit)),
    theta[c("p0","gamma","alpha","beta","isol_factor","eta")]
 ),
 est=c(
    "lambda.A", "lambda.B", "lambda.C", "lambda.D", "lambda.E", "lambda.out", "gamma"
 data=dat
) -> h
## Fit the model:
optim(
 par=log(coef(h)[
   c("lambda.A", "lambda.B", "lambda.C", "lambda.D", "lambda.E", "lambda.out", "gamma")
 ]),
 fn=h,
 control=list(reltol=1e-3)
) -> out
h(out$par)
coef(h)
```

simuldat

simuldat

Description

simuldat simulates data representing the flow of a body of patients through a hospital over a specified window of time.

simul_patient simulates a single patient's history of movement, testing, isolation, and infection.

Usage

```
simuldat(
  nbeds = c(50, 60),
  arrival = 20,
  window = c("1999-12-31T23:59:59+0000", "2003-01-01T00:00:00+0000"),
  units = list(A = list(shape = 10, scale = 2/10), B = list(shape = 5, scale = 0.5/5), C
  = list(shape = 1, scale = 0.5/1), D = list(shape = 1, scale = 3/1), E = list(shape = 0.2, scale = 8/0.2), out = list(shape = 0.5, scale = 300/0.5)),
  visits = list(size = 0.5, mu = 9),
  uperv = list(size = 0.5, mu = 9),
  uperv = list(size = 1, mu = 0.5),
  min_dur = 1/24,
  testing_freq = c(A = NA, B = NA, C = 1/7, D = 1/7, E = 1/7, out = NA),
  isolation = list(on = 1/50, off = 1/50),
```

```
infection = list(lambda = c(A = 0.01, B = 0.02, C = 0.001, D = 0.5, E = 0.1, out = 0.01, B = 0.01, B = 0.01, B = 0.01, B = 0.01, D = 0
                         0.05), gamma = 0.01, p0 = 0.1, isol_factor = 0.2),
            alpha = 0.02,
           beta = 0.1,
            verbose = getOption("verbose", TRUE)
)
simul_patient(
            patient,
             t0,
             tf,
            units,
            visits,
            uperv,
           min_dur,
             testing_freq,
             isolation,
             infection,
            alpha,
           beta
)
indep_infect(lambda, gamma, p0, isol_factor, times, loc, isol)
```

Arguments

nbeds	upper and lowe	er bounds on	number of beds

arrival Poisson arrival rate of new patients

window window of simulation

units a named list with one entry per unit. Each entry is itself a list with the parameters

of the Gamma-distribution for the duration of stay in the unit.

visits list containing parameters of a negative binomial distribution for the number of

visits per patient.

uperv list containing parameters for a negative binomial distribution for the number of

units visited per visit.

min_dur minimum duration of stay in any unit

 ${\tt testing_freq} \qquad {\tt named \ numeric \ vector \ of \ unit-specific \ testing \ frequencies}.$

isolation list containing parameters of the isolation model

infection parameters of the infection model

alpha, beta false positive and negative testing error rates

verbose run-time information?
patient patient name or number

t0, tf initial and final times of patient itinerary

lambda force of infection

gamma	recovery rate
p0	initial probability of infection
isol_factor	reduction in susceptibility due to isolation
times	times at which status is reported
loc	location of patient at each time
isol	isolation status

Details

Simulate hospital movement, testing, and isolation data.

Value

infection status vector

See Also

More on simulated data: fake_data

More on the independent infection model: independent

```
library(tidyverse)
library(lubridate)
## Examine the data:
fake_data
## Verify certain conditions hold:
stopifnot(
  `admission condition violation`=fake_data |>
   group_by(patient, visit) |> slice_head() |>
    filter(event!="admit") |> nrow()==0,
  `discharge condition violation`=fake_data |>
    group_by(patient,visit) |> slice_tail() |>
    filter(event!="discharge",event!="stop") |> nrow()==0,
  `unit violation`=fake_data |> filter(is.na(unit)) |> nrow()==0,
  `event violation`=fake_data |> filter(is.na(event)) |> nrow()==0
)
fake_data |>
  mutate(
   time=as.numeric(
      as.duration(
       interval(date, start="2000-01-01T00:00:00+0000")
      units="day"
  ) -> fake_data
```

```
fake_data |>
 group_by(patient, visit) |>
 summarize(dur=max(time)-min(time)) |>
 ungroup() |>
 group_by(patient) |>
 summarize(dur=sum(dur)) |>
 ggplot(aes(x=log10(dur)))+
 geom_histogram(bins=20)+
 labs(title="total duration of hospitalization")+
 theme_bw()
fake_data |>
 group_by(patient, visit) |>
 summarize(dur=max(time)-min(time)) |>
 ungroup() |>
 ggplot(aes(x=log10(dur)))+
 geom_histogram(bins=40)+
 labs(title="duration of hospital visit")+
 theme_bw()
fake_data |>
 filter(
   event!="test",
    event!="isolate",
   event!="release",
   event!="stop"
 group_by(patient, visit) |>
 arrange(time) |>
 mutate(dur=lead(time)-time) |>
 ungroup() |>
 filter(unit!="out",!is.na(dur)) |>
 ggplot(aes(x=log10(dur),fill=unit,group=unit))+
 geom_histogram(aes(y=after_stat(density)),bins=40)+
 facet_grid(unit~.,scales="free_y")+
 labs(title="duration of stay by unit")+
 theme_bw()
fake_data |>
 group_by(patient) |>
 summarize(ntest=sum(event=="test")) |>
 ungroup() |>
 ggplot(aes(x=ntest))+
 geom_histogram(binwidth=1,center=0)+
 labs(title="number of tests per patient")+
 theme_bw()
fake_data |>
 arrange(time) |>
 mutate(
   dn=case_when(
     event=="admit"~1L,
      event=="discharge"~-1L,
```

```
TRUE~0L
   ),
   occ=cumsum(dn)
 ) |>
 ggplot(aes(x=date,y=occ))+
 geom_step()+
 labs(title="hospital occupancy")+
 theme_bw()
fake_data |>
 arrange(time) |>
 select(date,test.result=result,isol,infected) |>
 pivot_longer(c(test.result,isol,infected)) |>
 filter(!is.na(value)) |>
 ggplot(aes(x=date,y=value,color=name))+
 geom_point()+
 geom_smooth()+
 guides(color="none")+
    title="infection and isolation status, test results",
   y=""
 )+
 facet_grid(name~.)+
 theme_bw()
fake_data |>
 filter(event=="test") |>
 mutate(
   interval=cut(time,breaks=72,ordered=TRUE)
 ) |>
 select(interval,time,infected,isol,result) |>
 pivot_longer(c(infected,isol,result)) |>
 group_by(name,interval) |>
 summarize(
   time=mean(time),
   prev=mean(value),
   n=n()
 ) |>
 ungroup() |>
 ggplot(aes(x=time,y=prev,group=name,fill=name))+
 geom_col(position="dodge")+
 labs(title="infection, isolation, and detection through time")+
 theme_bw()+
 theme(axis.text.x=element_text(angle=90))
fake_data |>
 filter(event=="test") |>
 select(infected,result) |>
 count(infected,result) |>
 group_by(infected) |>
 mutate(prob=n/sum(n)) |>
 ungroup()
```

14 stobfun

stobfun

Stateful objective functions

Description

Convenience functions for constructing and working with stateful objective functions ('stobfun'-class objects).

Usage

```
stobfun(
  embed,
  params,
  est = character(0),
  log = character(0),
  objfun,
  data,
  ...
)

filterfun(embed, params, filtfun, data, ...)

transf_fns(log = character(0), logit = character(0), est = character(0))

embedding(...)

## S3 method for class 'stobfun'
coef(object, ...)
```

embedding (see embedding).

Arguments

embed

params	vector of parameters
est	character: names of parameters to be estimated
log	character; names of parameters to log transform.
logit	character; names of parameters to logit transform.
objfun	underlying objective function
data	data
•••	When furnished to stobfun, additional arguments are passed to objfun. When furnished to embedding, arguments define th embedding. When furnished to coef, additional arguments are ignored.
filtfun	the function that actually applies the filter
object	'stobfun'-class stateful objective function

transmission 15

Details

objfun will be called as objfun(theta,data), where theta is the nested list constructed according to the given specifications and data is the data.

Value

transf_fns returns a list of two functions. The first is the transformation to the estimation scale; the second is its inverse.

embedding returns the embedding function corresponding to the given specification.

coef(f) returns the parameter vector corresponding to the last call of the stateful objective function f.

Construction and usage of stateful objective functions

A stateful objective function is an ordinary function that can be used as an objective function in an optimization problem. In particular, it can be passed to optimizers such as optim, subplex, or nloptr. It is stateful in the sense that it remembers the argument with which it was last called.

To construct a stateful objective function, call the constructor function for the model of interest. The constructor function requires that you pass a vector of model parameters: this gives the default parameter values. It also requires that you pass the data and the names of the parameters that you wish to estimate. The constructor will return an object of class 'stobfun'.

Having constructed a 'stobfun' stateful objective function, you can pass this to any suitable optimizer. Once the optimizer has returned, it is important that you call the function one last time, at the parameters the optimizer has returned (see examples). This ensures that the stored parameters are those at the (putative) optimum. You can retrieve these parameters via a call to coef.

See Also

More on stateful objective functions: independent, transmission

transmission

Transmission model

Description

Model under which the force of infection in each unit is proportional to the prevalence of infection in that unit.

Usage

```
trans_homog_filter(params, data, tol = 1e-04, maxit = 10)
trans_homog_objfun(params, data, est = character(0), tol = 1e-04, maxit = 10)
trans_unit_spec_filter(params, data, tol = 1e-04, maxit = 10)
```

16 transmission

```
trans_unit_spec_objfun(
  params,
  data,
  est = character(0),
  tol = 1e-04,
  maxit = 10
)
```

Arguments

params named vector of parameters

data patient movement, isolation, and testing data

tol positive scalar; convergence tolerance (mean difference).

maxit scalar integer; maximum number of fixed-point iterations. If tol is not achieved

in maxit or fewer iterations, an error is generated.

est names of parameters to estimate

Details

The basic transmission model assumes that the force of infection on an unisolated patient is

$$\lambda = b(P_u + aP_i) + \iota,$$

where P_u , P_i are the prevalences among unisolated and isolated patients, respectively and ι represents the risk of acquisition unrelated to local prevalence. On an isolated patient, this force of infection is reduced by the factor isol_factor.

Parameters in the model include:

b transmission rate. In the homogeneous model, this is constant across units in the hospital; in the unit-specific model, there is one value of b for each unit.

lambda.out force of colonization outside the hospital

iota baseline force of colonization

gamma recovery rate

alpha,beta false positive and negative probabilities

p0 probability of infection on admission

a multiplicative effect of contact isolation on transmissibility

isol_factor multiplicative effect of contact isolation on susceptibility

eta additional information on infection status

trans_homog_filter runs a fixed-point Bernoulli filter for the transmission model with global b and gamma.

trans_homog_objfun is a stateful objective function for the transmission model with global b and gamma.

trans_unit_spec_filter runs a fixed-point Bernoulli filter for the transmission model with unit-specific b.

trans_unit_spec_objfun is a stateful objective function for the transmission model with unitspecific transmission rates, an out-of-hospital force of infection parameter, and recovery rates that can be different inside and outside of hospital. transmission 17

See Also

More on stateful objective functions: independent, stobfun()

```
library(tidyverse)
library(hte)
set.seed(339613584)
fake_data |>
  ## filter out tests prior to day 250
  filter(
    event!="test" | time < 250,
   time < 1000
  ) |>
  select(-infected) |>
  arrange(patient,time) -> dat
trans_homog_objfun(
  params=c(
   a=1,b=0.1,gamma=0.01,p0=0.1,lambda.out=0.01,
   iota=0,isol_factor=0.2,alpha=0.02,beta=0.1,eta=0.5
  ),
  est=c("a","b"),
  data=dat
) -> f
f(\log(c(0.05,0.4)))
coef(f)
optim(
  par=log(c(0.5,0.4)),
  fn=f,
  control=list(reltol=1e-2)
) -> out
f(out$par)
coef(f)
trans_homog_filter(params=coef(f),data=dat) -> ff
ff |> filter(logLik!=0)
trans_unit_spec_objfun(
  params=c(
   b.A=0.1,b.B=0.1,b.C=0.1,b.D=0.1,b.E=0.1,
    gamma=0.01,p0=0.1,
    lambda.out=0.1,
    iota=0,
   isol_factor=0.2,
   alpha=0.02,beta=0.1,
   eta=0.5
  ),
```

18 twostate

```
data=dat
) -> f
f()
coef(f)
```

twostate

two state

Description

Two-state Markov process with on and off rates

Usage

```
twostate(on, off, tf, t0 = 0)
```

Arguments

on, off on and off rates t0, tf initial and final times

Index

```
* independent model
    independent, 7
    simuldat. 9
* simulated data
    fake_data, 4
    simuldat, 9
* stateful objective functions
    independent, 7
    stobfun, 14
    transmission, 15
* transmission model
    transmission, 15
Bernoulli_filter, 2
Bfilter (Bernoulli_filter), 2
coal_last, 3
coef, 14
coef, stobfun-method (stobfun), 14
coef.stobfun(stobfun), 14
embedding, 14
embedding (stobfun), 14
fake_data, 4, 11
filterfun (stobfun), 14
hte (hte-package), 1
hte, package (hte-package), 1
hte-package, 1
indep_homog_filter(independent), 7
indep_homog_objfun (independent), 7
indep_infect (simuldat), 9
indep_unit_spec_filter (independent), 7
indep_unit_spec_objfun(independent), 7
independent, 7, 11, 15, 17
nloptr, 15
optim, 15
```

```
simul_patient (simuldat), 9
simuldat, 4, 8, 9
stobfun, 8, 14, 17
subplex, 15

tibble, 2
trans_homog_filter (transmission), 15
trans_homog_objfun (transmission), 15
trans_unit_spec_filter (transmission), 15
trans_unit_spec_objfun (transmission), 15
transf_fns (stobfun), 14
transmission, 8, 15, 15
twostate, 18
```